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GEOSTATISTICAL MODELLING OF MICROFOSSIL ABUNDANCE  
DATA IN UPPER JURASSIC SHALE, TOJEIRA SECTIONS,  
CENTRAL PORTUGAL

by

Kazim Nazli

A thesis submitted to the School of Graduate Studies in partial  
fulfillment of the requirements for the degree of M. Sc. in Geology

UNIVERSITY OF OTTAWA

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# ABSTRACT

Every count can be seen as a statistical sample from a statistical population in paleontology. It can be defined as the collection of individuals of a specific group of microfossils that is present in the geological sample from which the count comes. In general, when sampling from a stratigraphic section one should consider only the vicinity within the layer from which the sample comes.

The sampled population proportions ( percents ) can be considered to be random variables.

Along a stratigraphical column the proportions of a certain taxon may show an overall increase or an overall decrease; these phenomena are called trends. There may also exist more local increases alternating with decreases. These systematic changes ( fluctuations ) may be expressed by the autocorrelation coefficient value between successive proportions for a number of the samples ordered from bottom to top.

Care is required in evaluating such autocorrelation coefficient values, because one of the series of variables, the sample numbers, has a fixed, logical ( stratigraphic ) order. This logical order of the samples may be correlated to the other series of variables, the proportions.

There may be an interdependence in the sense that some taxon property ( e.g., its proportion ) in a certain sample is strongly related to the one in the sample immediately below. Series with interdependence because of an inherent logical order in time are called time series.

The analysis of the time series is helped by geostatistical modelling

based on probability theory. Probability models give a good description of reality, when uncertainty is a factor.

This method of treating autocorrelated proportions used for this study was originally developed by F.P. Agterberg. The geological applications of the geostatistical model used for this project have previously been restricted to ore minerals in ore deposits and occurrence of volcanic rocks in large regions ( Agterberg, 1974 and 1984 ).

It is useful in spatial pattern analysis to assume that a feature can be significantly correlated with features that occur at some distance from it. A basic tool of spatial pattern analysis is the autocorrelation function of a random variable. The method to be discussed in this research can be applied to phenomena that occur in either two - dimensional or in three dimensional space.

When the geostatistical model was applied, it was required to estimate R values representing correlation coefficients between proportion variables for volumes and a binary variable defined for presence or absence of the feature for which the proportions were determined. These R values can be derived from the mean and variance of taxon proportion distributions by using the FF FORTRAN program that solves Hermite polynomials. Alternatively, the R - values can be determined from graphs or tables based on output from the FF FORTRAN program. New graphs and tables, especially for small mean proportions and their variances, have been prepared for this thesis project.

Several models in one or two dimensions produce exponential autocorrelation functions ( Matern ,1960 and Switzer,1965 ) which are frequently used in one - dimensional time series analysis. When the autocorrelation function is known, it is possible to estimate the variance of the distribution of a taxon measured as a proportion in a sample derived from a volume of any size and shape.

A model that can often be used is the signal - plus - noise model of the statistical theory of communication. The signal may satisfy an autoregressive process model. The noise ( random shock) consists of uncorrelated values and is independent of the signal.

Theoretical models for frequency distributions in spatial analysis have previously been proposed by several authors. These include the Neyman - Scott clustering model, the de Wijsian model, the Prokhorov model and a model based on the transfer function theory of Matheron which is the basis for the model used in this project.

When a random variable has both a stationary mean and a covariance function, it may be assumed that the discrete Gaussian model of Matheron can provide useful methods to obtain frequency distribution models in spatial analysis.

Before the start of the geostatistical modelling of the frequency distribution of the microfossils, the time series were tested as to whether or not, they could be described by various autoregressive process models that would explain the data set.

Newly prepared FORTRAN as well as existing SAS package programs ( ARIMA, SPECTRA and CORR procedures ) were used for

statistical calculations and SAS graphic software was used for the plotting of graphs.

For application of the geostatistical modelling, two stratigraphic sections were studied. They are called Tojeira 1 and Tojeira 2 and are located in the Montejunto area of central Portugal. Data were taken from a 1986 Ph. D. thesis by B. Stam and also provided by F.M. Gradstein who identified and counted species of foraminifers from samples newly collected during the summer of 1986. With Stam's data, 3 of 14 species in the Tojeira 1 section and 5 of 14 species in the Tojeira 2 section were shown to have exponential autocorrelation functions. They are Eoguttulina sp., Epistomina mosquensis and Ophthalmidium strumosum in Tojeira 1 and Eoguttulina sp., Epistomina mosquensis, Ophthalmidium strumosum, Spirillina tenuissima and agglutinants in the Tojeira 2 section.

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## CHAPTER 1: Introduction

### 1.1 Purpose of study

The objective of this new research project is the development of techniques to establish the reliability of microfossil abundance data in micropaleontology.

Statistical models are applied for estimating the precision of counts of species of different types. These methods will allow estimation of the precision of averages computed from values for a single taxon in samples of different size. It also has a bearing on the correlation of data on abundance of many taxa with one another and in relation with lithological information.

This methodology can be used in several other fields of geoscience. These include chemical analysis of rock samples, petrographic modal analysis, geostatistical ore - reserve estimation, geochemical exploration, line - spacing problems in exploration geophysics, drilling for hidden targets, and regional resource estimation.

The methods used are applicable to discrete or continuous variables defined for samples located in two dimensional or three - dimensional space. Spatial pattern analysis is an important topic of study in geomathematics. Generally, it is useful in spatial pattern analysis to assume that a feature can be significantly correlated with features which occur at some distance from it. For this reason, geological pattern analysis may require a more elaborate mathematical

apparatus than is available in current methods of pattern recognition, to perform the multivariate analysis of attributes observed simultaneously at the same points in space.

Application of this methodology to the bivariate normal distribution leads to the "prob - normal" distribution which has turned out to be very useful. A similar type of distribution can be obtained by applying the method to a bivariate exponential distribution (Agterberg 1984).

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## 1.3 Geology

### 1.3.1 Introduction

The data used in this thesis were derived from published reports and from information collected by the author following field investigation in 1986 of several geological sections.

The relevant sections were chosen carefully. The first consideration was that they had to be well exposed for detailed sampling. Secondly, the sections had to fit into a well known micropaleontologic and biostratigraphic framework ( Age - control ). Thirdly, some aspects of the paleogeographic setting of the areas had to be known. This was necessary to enable recognition of trends in basin tectonics and basin development. The Montejunto area of central Portugal, in the Lusitanian basin, was found to fit these criteria.

Much of the literature concerning the stratigraphy of Portugal has been written in French and Portuguese, and published in journals and memoirs that are not readily accessible.

The biostratigraphic framework, based on ammonites, is well known and discussed by Choffat, ( 1880 - 1893 ), Ruget - Perrot ( 1961 ), Mouterde et al. ( 1972 ), and Poulton ( 1983 ).

The paleogeographic setting of the areas has been documented by Ruget Perrot ( 1961 ), Mouterde and Ruget ( 1975 ), Wilson ( 1975a ), Wilson and Exton ( 1979 ) and Exton and Gradstein ( 1984 ). Stam ( 1986 ) has made detailed comparisons between Middle and Late Jurassic foraminifera from Portugal and the Grand Banks of Newfoundland. He has used quantitative methods to examine some paleoecological parameters controlling the distribution and abundance of Jurassic benthonic foraminifera. He also evaluated the biostratigraphic

## Usefulness of Jurassic planktonic and benthonic foraminifera.

### 1.3.2 Paleogeography of the Lusitanian basin

A useful summary of the stratigraphy of the Lusitanian basin can be found in the guidebook of the field trips accompanying the 27<sup>e</sup> I.G.C. (Paris), Livret Guide G - 14 ( 1980 ). Data on Triassic exposures were presented by Palain ( 1976 ) and the Lower Jurassic has been summarized by Hallam ( 1971 ) and Mouterde and Ruget ( 1975 ). The Triassic and Jurassic rocks of Portugal have been studied since the end of the nineteenth century ( Choffat 1880 ). The Lusitanian basin originated in the Late Triassic - Early Jurassic as a result of movements along Hercynian ( Paleozoic ) normal basement faults; one of which is the Nazare strike - slip fault. Triassic sediments in the Lusitanian basin rest unconformably on igneous, metamorphic and sedimentary rocks of Precambrian and Paleozoic ages ( Da Costa, 1950; Howie and Barss, 1975 ). The Late Triassic through Middle Jurassic development of the basin was generally uniform. It started with the deposition of nonmarine and marginal marine to shallow marine sands, shales and evaporites. Dolomites and dolomitic limestone were deposited throughout the Sinemurian ( Doubinger et al. 1970 ; Mouterde et al. 1973; Mouterde and Ruget, 1975 ). This was followed in the late Sinemurian and throughout the Pliensbachian by normal marine sedimentation of shales, alternating with limestone. These late Sinemurian sediments contained the first occurrence of ammonites. At

the beginning of the Toarcian. the shale facies became predominant. with an upward trend towards more calcareous sedimentation. This trend continued throughout the Aalenian and Bajocian until the Bathonian. At this time most of the basin was covered by shallow carbonate deposits. During the Callovian, also characterized by shallow water deposits, reactivation of basement faults and halokinesis changed the basin configuration to such an extent that a northern and a southern sub - basin can be recognized, separated along the Nazare fault line ( Stam, 1986 ).

In the northern sub - basin, the deposition of oolitic limestone and sandstone over shaly and " reefal " limestone ( Mouterde et al., 1973 ), indicates a ( middle ) Callovian regression. lower Oxfordian sediments appear absent, because of uplift ( Wilson and Exton, 1979 ). Sedimentation resumed in the late Oxfordian with the deposition of marginal marine carbonates and fresh water sediments forming algal marsh facies ( Wilson, 1975 ). This was followed in the Kimmeridgian by fluvial clastics. In late early Oxfordian times, more rapid subsidence in the southern sub - basin resulted in normal marine sedimentation with abundant ammonites. Sediments include high energy carbonates ( Wilson, 1975 ). In late Oxfordian time this high energy facies was gradually replaced by low - energy shale sedimentation and by the end of the Oxfordian, and at the beginning of the Kimmeridgian, shale facies were predominant. Reactivation of basement faults and salt movements in the early Kimmeridgian resulted in a complex facies distribution, with marine clastics in the Montejunto area, and shelf

carbonates to the northeast of the Montejunto area.

### 1.3.3 Montejunto area

The Montejunto area is located some 50 km north of Lisbon ( Fig.1.1 ). and is dominated by the 664 m. high "Montejunto" mountain some 7 km. northeast of Vila Verda.

An almost complete, but tilted sequence of Bathonian through Kimmeridgian marine sediments, rest on the top and flanks of Montejunto.

The geology and macropaleontology of the area has been described by Choffat ( 1880 - 1893 ), Ruget - Perrot ( 1961 ), and Mouterde et al. ( 1972, 1973 ).

The formations found in this sequence are the Cabacos, Montejunto, Tojeira, Cabrito, Abadia, and Amaral Formations ( Figs. 1.2 - 1.3 ).

## 1.4 Stratigraphy

The stratigraphy of the area has been explained by Stam ( 1986 ) as follows:

### 1.4.1 Middle Jurassic

Rocks of the middle Jurassic age are exposed near the top of the Montejunto mountain. The oldest strata consist of thick bedded ( 0.5 -

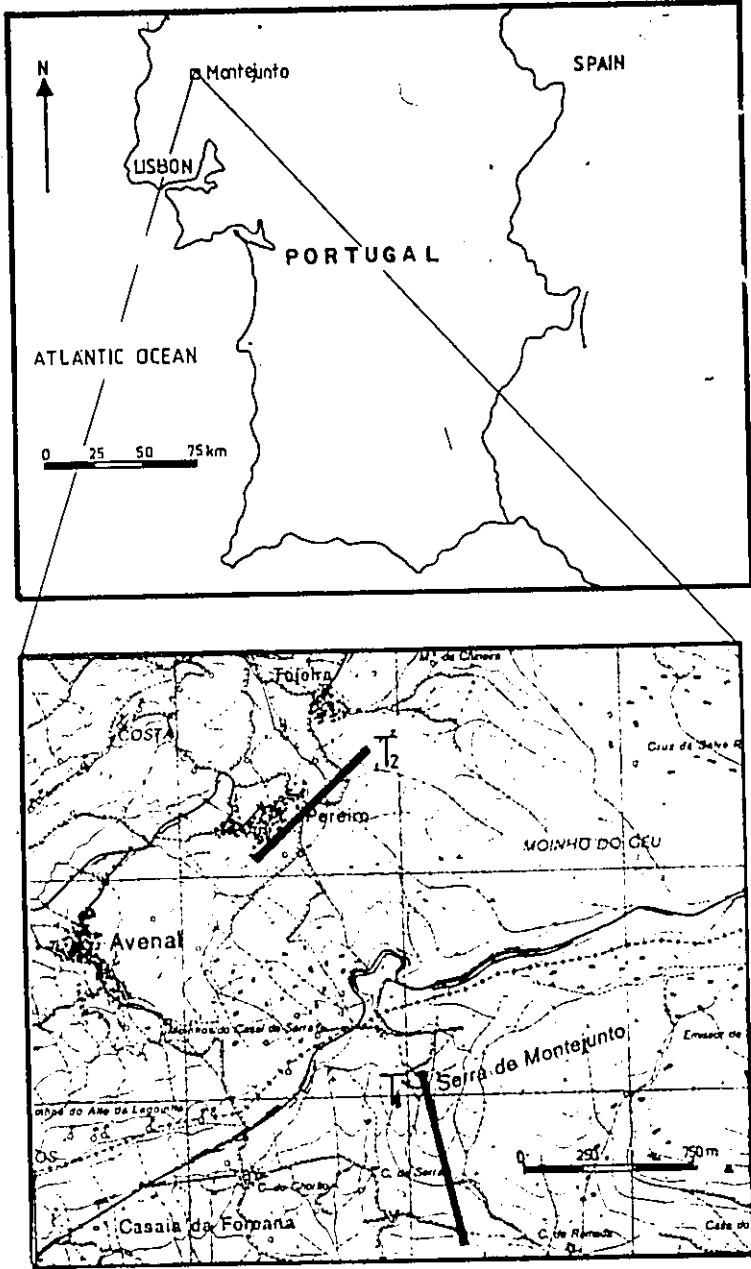


Figure 1.1 . Location map of the Montejunto area Tojeira 1 and Tojeira 2 sections

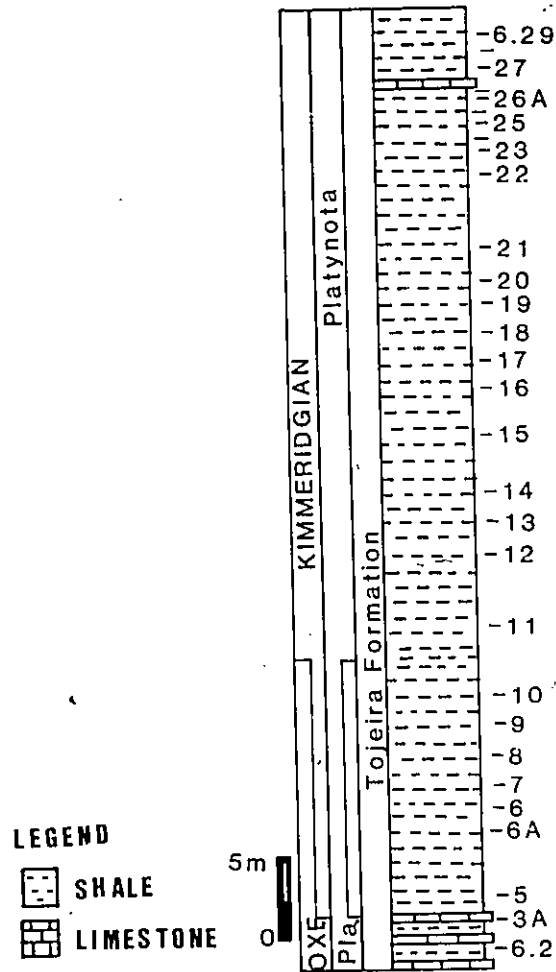


Figure 1.2 The Tojeira 1 section with sample numbers 6.2 - 6.29 ( after Stam, 1986 )

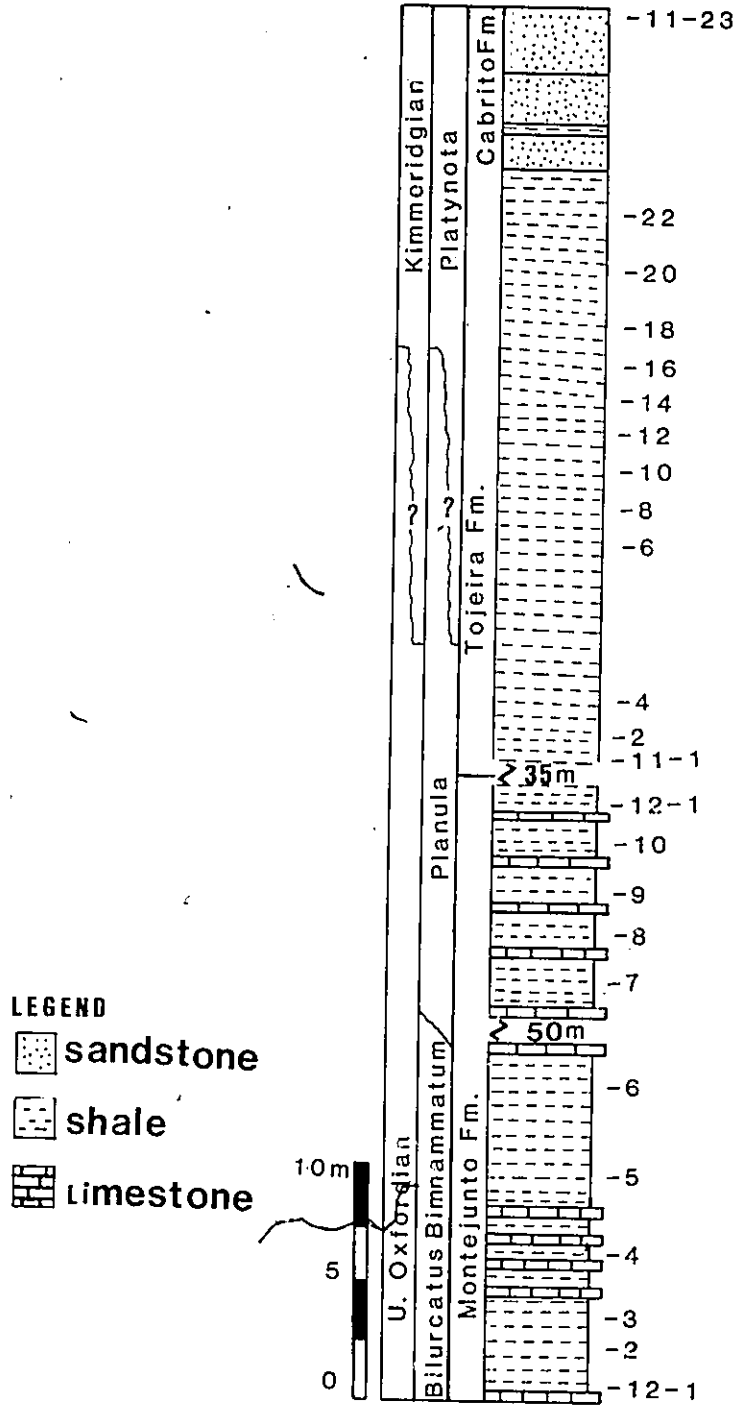


Figure 1.3 The Tojeira 2 section with sample numbers 12.1 - 12.11 and 11.1 - 11.23 ( after Stam 1986 )

2m ) beige to grey, mainly micritic, sometimes oolitic/pisolitic limestones with rare crinoids and *Ostrea*. No ammonites were observed nor are any reported in the literature. Based on occurrences of the larger foraminifer *Meyendorffina bathonica* Bizon, Zbyszewski et al., ( 1966 ) believe these strata belong in the upper Bathonian.

The Upper Jurassic formations are separated from these Middle Jurassic sediments by a hiatus. The upper Callovian Lamberti zone and the lower Oxfordian Mariae zone are missing.

#### 1.4.2 Upper Jurassic

##### 1.4.2.1 Cabacos Formation ( Cordatum - Transversarium zones )

The basal 30 m of this formation consists of highly bituminous platy, thinly bedded, grey and shaly limestone. This is overlain by at least 200m of thickly bedded ( 0.5 - 2m ) yellow to dark brown bituminous limestone, which is fine - grained, micritic and locally oolitic. Ammonites occur throughout the formation but are rare in the lower part.

##### 1.4.2.2 Montejunto Formation ( Transversarium - Planula zones )

The lower part of this 150 to 200 m thick formation consists of thick bedded ( up to 2m ) white to grey micritic limestone, changing to a thickly bedded alternation of bluish - grey limestone and hard grey shale. Ammonites are common throughout the formation. At the top

near the ( covered ) contact with the overlying Tojeira Formation.  
several shale and limestone beds are very rich in ammonites.

#### 1.4.2.3 *Tojeira Formation* ( Planula - Platynota zones )

This formation is at least 70 m thick and consists almost entirely of dark grey shales, in which limonite concretions are common. Some calcareous shale intercalations occur in the lower part of the formation. Locally, in the middle part of the formation, the shale is brownish - red. Close to the boundary with the overlying Cabrito Formation, there is a marked increase in silt content. Pyritized ammonites are common to abundant in the lower part, but are relatively rare in the upper part of formation.

#### 1.4.2.4 *Cabrito Formation* ( ? Platynota zone )

This formation consists of a massive, coarse - grained, micaceous sandstone unit and a conglomeratic unit containing pebbles of quartz and pink feldspar. The sandstone is dark yellow to brown in colour, near - massive, and is generally poorly sorted. Crinoids, shell debris and large ( up to 40 cm ) fragments of coral colonies are common to abundant. In the Montejunto area, the Cabrito formation is about 40 - 50 m thick, thickening rapidly towards the west. This was interpreted by Mouterde et al. ( 1973 ), as indicative of a westerly source for the influx of these clastics. A few badly preserved ammonites have been reported

(Mempel, 1955 : Mouterde et al., 1973 ).

#### 1.4.2.5 *Abadia Formation* ( ? Kimmeridgian )

Characteristic of this formation are the blue colour of its sediments and the presence of abundant black plant remains. Ostracods and thin-shelled bivalves are common, but no ammonites were observed. It consists of very thickly bedded ( up to 4 m ), coarse-grained blue coloured sandstone that shows ripple marks and channelling. The sandstone alternates with bluish sandy siltstone/shale.

#### 1.4.2.6 *Amaral Formation* ( ? Kimmeridgian )

This formation is up to 70 m thick, and consists of well sorted brownish quartz sandstone containing plant remains, and sometimes showing cross bedding. A few *Ostrea* and pelecypods are present, but no ammonites were observed.

### 1.5 Section descriptions

Two sections have been sampled in detail by the author during the summer of 1986. The first, called Tojeira 1, is located near the hamlet of Ramada, and covers the top 65m of the Tojeira Formation. The second, Tojeira 2, is located near the villages of Pereiro and Tojeiro and covers the top of the Montejunto Formation and the upper part of

the Tojeira Formation. Stam ( 1986 ) has made a quantitative analysis of the foraminiferal assemblages in 137 samples from five Portuguese sections of Middle and Late Jurassic age, two of which were sections Tojeira 1 and Tojeira 2 ( Figs.1.4 - 1.5 ). Sampling strategy involved trying to match as many levels as possible to those previously sampled by Stam ( 1986 ) and achieving regular spacing of the samples. Although there is no certainty that the sedimentation rate was constant, the monotonous Tojeira shale lithology suggests so. Less continuous outcrops in the Tojeira 2 section did not allow regular sample intervals. Samples T<sub>2</sub>/1 to T<sub>2</sub>/3 are from the Tojeira 2 section and were taken from a small outcrop of the Montejunto Formation in the down faulted block west of Montejunto Mountain. Sample T<sub>2</sub>/4 was taken along a dirt road halfway between T<sub>2</sub>/3 and T<sub>2</sub>/5. These samples are separated by a gap in the outcrop of approximately 50 m. Samples T<sub>2</sub>/5 and T<sub>2</sub>/6 were taken along the asphalt road that runs from Pereiro to Tojeira and correspond to the top of section 12 of Stam ( 1986 ). Samples T<sub>2</sub>/1 - 5 could be matched in the field to the approximate lithostratigraphic position ( ± 1/2m ) of samples 12/2 - 7 studied by Stam ( 1986 ) ( Fig.1.3 ). T<sub>2</sub>/7 is a new sample taken from a ditch halfway between T<sub>2</sub>/6 and T<sub>2</sub>/8. Stam (1986 ), had previously indicated that there was no exposure at this point. This sample location was probably estimated too low in the section, and should be estimated at an 80 m. thickness. Samples T<sub>2</sub>/8 and T<sub>2</sub>/9 are on the W side of the asphalt road just before the first house entry from Pereiro to Tojeira. Samples T<sub>2</sub>/10 to T<sub>2</sub>/13 were

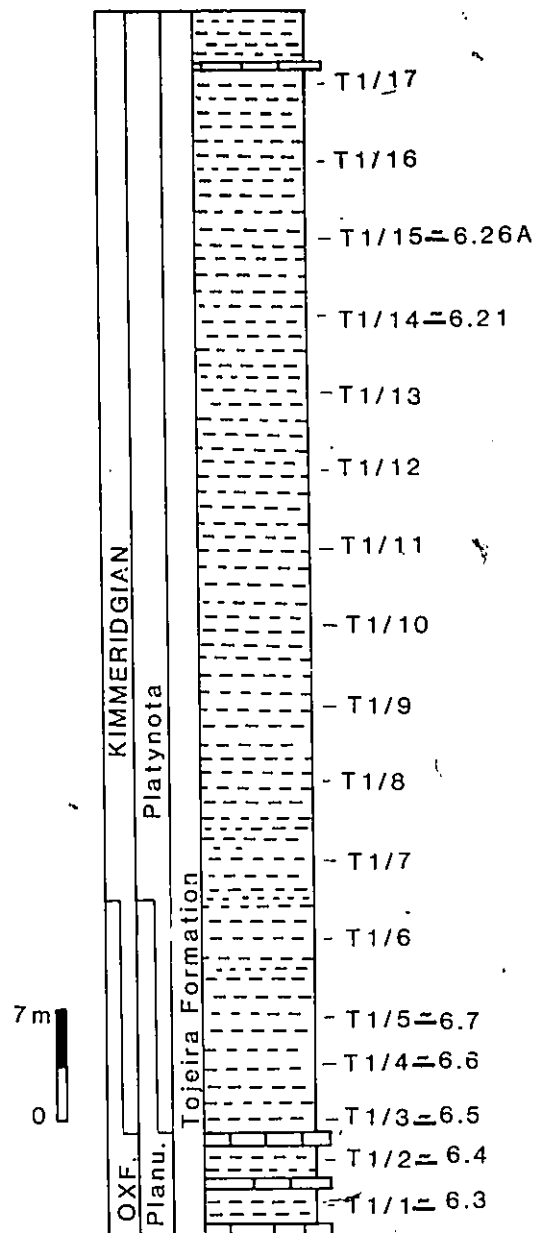


Figure 1.4 The Tojeira 1 section with F.M. Gradstein's sample numbers T1/1 - T1/17 as they correspond to Stam's samples

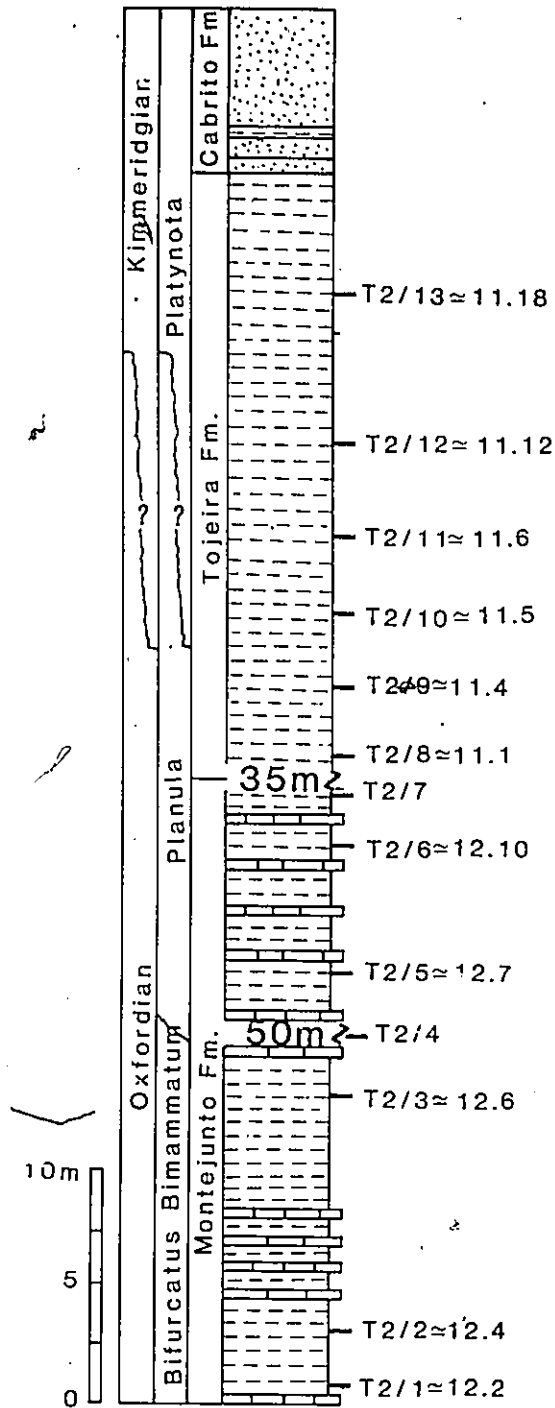


Figure 1.5 The Tojeira 2 section with F.M. Gradstein's sample numbers T2/1 - T2/13 as they correspond to Stam's samples

taken from a steep outcrop above the dirt road just West of the asphalt road. The thickness of this outcrop mentioned by Stam ( 1986 ), was probably underestimated by 30 % or more. Because of easily recognizable landmarks in the section, all but a few of our samples could be matched to Stam's ( 1986 ) sampling levels, ( see Fig. 1.5 ).

Sample T<sub>1</sub>/1, from the Tojeira 1 section east of Montejunto, was taken as a first sample from under outcrop 17 of Stam ( 1986 ).

Samples T<sub>1</sub>/2, T<sub>1</sub>/2A, 2B, 2C, and 2D were collected, at 5m intervals. in the same layer located above-outcrop 17. Samples T<sub>1</sub>/3 and T<sub>1</sub>/17 were taken in order at 5 m intervals. Several of the newly collected Tojeira 1 samples correspond with Stam's ( 1986 ) samples, as shown in Fig. 1.4.

#### 1.6 Preparation of Data

The samples were prepared in the laboratories of the Geological Survey of Canada, Bedford Institute of Oceanography, Dartmouth, Nova Scotia. Samples were processed according to the " Industrial Soap Method " of Thomas ( 1985, 1986 ). This included washing the samples over a set of sieves sized 500, 125, and 63 microns. The 63 - 125 micron size fraction was used neither for this study nor by Stam (1986). All residues were split such that several hundred foraminiferal shells remained in one part. Each split was spread over one or more picking trays and picked clean of foraminifers. Next, using a microsplitter the benthic taxa were identified and counted by F. Gradstein ( in 1987 ).

Percentages of taxa were calculated and presented as shown in Tables 1.3 - 1.4. Stam ( 1986 ) previously used the same method and his results are shown in Tables 1.1 - 1.2. A planktonic/ benthonic ( P/B ) ratio was also calculated and shown for the data sets in both tables. Faunal analysis of the Tojeira 1 and Tojeira 2 sections are discussed below. Stam ( 1986 ) took thirty one samples for foraminiferal analysis. The frequency patterns of seventeen taxonomic categories were examined, fifteen of which concern proportion counts on species or species groups. These taxonomic units are: Eoguttulina sp. ( E. oolithica and E. metensis ), Epistomina mosquensis, E. uhligi, all other Epistomina sp., Lenticulina muensteri , all other Lenticulina sp., Nodosaria/Dentalina sp., Pseudolamarckina rjasanensis , Spirillina elongata, Spirillina infima, Spirillina tenuissima, Ophthalmidium carinatum, Ophthalmidium strumosum, agglutinants ( Ammobaculites sp. Reophax sp., Trochammina sp., Textularia sp., Bigenerina sp.). All other species are grouped in a Restgroup; usually comprising less than 15% of the benthonic microfauna.

For this study, samples used to create a geostatistical model included the thirty one samples reported by Stam ( 1986 ) plus fifteen samples collected by the author from the Tojeira 1 and thirteen new samples collected by the author and the thirty samples reported by Stam ( 1986 ) for the Tojeira 2 section.

TABLE 1.1 DATA SET OF TOJEIRA 1 SECTION

Sample number	A1	A2	A3	A4	A5	A6	A7	A8	A9	A10	A11	A12	A13	A14	15	16	17	18
6.2	5.0	4.1	.0	.0	1.5	7.9	6.6	4.8	10.4	.0	.0	1.8	.9	45.3	11.7	-	36	558
6.3	2.5	14.8	.0	.0	.0	24.7	5.7	4.9	.2	.0	.0	.9	18.8	17.9	9.6	.01	24	453
6.3A	.5	19.0	.0	.5	.0	22.5	3.8	4.6	1.1	.0	.0	.0	26.8	13.6	7.6	-	22	369
6.4	5.0	5.4	.2	1.2	.0	19.8	10.6	3.4	4.2	.0	.0	2.4	.4	26.1	21.3	-	34	501
6.5	2.6	1.8	15.1	.0	13.3	4.7	4.9	.0	.0	.0	.0	2.9	7.2	34.8	11.9	.01	32	511
6.6A	1.2	13.5	.84	1.5	16.2	5.4	.3	.6	14.4	.0	.0	3.8	6.0	11.7	7.0	.09	21	333
6.6	2.3	.22	24.2	11.8	.0	9.9	4.8	2.5	3.8	.2	.0	10.8	3.8	9.1	14.8	.02	30	372
6.7	5.4	3.4	18.1	3.4	.0	11.1	3.1	3.8	3.8	.2	.0	6.9	.6	29.2	11.0	.09	32	524
6.8	4.5	.3	12.4	1.0	.0	5.6	2.1	2.0	35.4	.0	5.1	5.3	.0	15.4	10.9	.13	26	396
6.9	6.4	2.6	26.1	4.3	.3	4.6	4.1	2.6	3.7	.0	10.9	5.7	.0	23.2	5.5	.40	27	349
6.10	5.1	1.1	27.6	2.5	2.5	1.4	.6	2.2	1.1	.3	29.6	1.7	.3	20.7	3.3	1.19	27	362
6.11	3.7	1.1	13.2	17.6	.0	23.0	6.1	2.0	.5	.2	10.9	1.1	.2	11.1	9.3	.08	31	614
6.12	21.6	1.3	23.6	3.0	3.0	9.3	3.3	.0	.0	.0	12.3	1.3	.0	15.0	6.3	.17	27	301
6.13	3.9	.6	23.3	5.7	4.8	8.7	2.7	.3	1.2	.0	14.0	1.5	.0	23.0	10.3	.60	26	335
6.14	7.6	1.5	15.5	.7	.9	4.8	3.8	1.8	2.4	2.3	4.8	5.0	.7	31.3	6.9	.44	37	459
6.15	2.4	.3	35.2	7.8	6.0	8.1	7.2	1.5	.6	.3	10.5	1.5	.0	15.1	3.5	.41	30	332
6.16	7.8	.5	21.4	3.1	7.7	15.2	5.5	1.4	2.1	.2	15.0	1.7	.0	11.2	4.5	.28	37	420
6.17	2.9	1.1	46.3	2.0	1.9	5.1	2.8	1.1	1.4	8.3	6.4	4.1	.0	13.2	3.4	.16	30	916
6.18	14.0	.3	18.0	1.7	5.3	3.9	7.2	2.8	.8	6.6	8.0	4.7	.0	20.8	5.9	.44	34	361
6.19	8.5	.8	39.7	.6	3.5	3.5	4.4	2.5	1.0	1.9	11.6	4.1	.0	8.3	9.6	.11	30	484
6.20	6.0	.8	30.4	6.0	1.4	9.1	7.4	1.5	1.9	.4	12.6	4.4	1.0	9.1	8.0	.06	35	517
6.21	11.4	.8	13.5	1.7	1.2	15.0	11.3	1.9	1.9	7.2	9.3	4.2	.0	13.5	7.1	.06	37	527
6.22	4.6	3.5	23.0	.3	.8	11.8	6.6	1.6	.5	2.4	11.8	7.6	.0	22.3	3.2	.30	35	382
6.23	3.3	1.5	33.5	.8	6.2	11.4	9.3	2.1	1.6	1.9	10.9	4.8	.0	7.7	5.0	.16	25	376
6.24	7.1	.6	37.1	3.2	3.1	6.6	3.6	1.5	1.2	5.8	1.0	3.4	.0	9.0	16.8	.11	35	412
6.25	7.4	.6	36.6	7.4	8.4	9.1	1.2	3.9	.6	1.9	.0	1.9	.0	13.6	7.4	.31	33	309
6.26	17.1	.9	38.8	4.9	.6	6.7	2.7	1.7	.6	2.9	3.2	.6	.0	13.3	6.0	.08	27	345
6.26A	2.2	1.0	34.8	3.7	3.4	16.2	3.3	2.5	.3	5.2	9.1	4.6	.0	7.0	6.7	.02	43	767
6.27	10.6	.3	36.5	1.2	1.8	11.9	8.8	.3	1.2	2.7	6.4	5.5	.0	8.8	4.0	.34	28	329
6.28	11.8	.0	27.4	3.4	2.3	14.0	2.4	2.0	.3	1.1	17.1	2.3	.0	9.4	6.5	.70	28	351
6.29	6.2	.2	16.7	3.2	2.5	6.5	.6	2.5	.0	5.3	21.6	4.9	.4	16.0	13.4	.64	29	474

17 = number of species  
18 = total number

A13 = S. infima  
A14 = S. tenuissina  
15 = Restgroup  
16 = P/B-ratio

A9 = O. carinatum  
A10 = O. sirumosum  
A11 = P. fisanensis  
A12 = S. elongata

A5 = Epistomina spp.  
A6 = L. muensteri  
A7 = Lenticulina spp.  
A8 = Nodosaria/Dentalina spp.

A1 = Agglutinants  
A2 = Ecgyttulina spp.  
A3 = E. mosquensis  
A4 = E. ubiqi

TABLE 1.2 DATA SET OF TOJEIRA 2 SECTION

Sample number	B1	B2	B3	B4	B5	B6	B7	B8	B9	B10	B11	B12	B13	B14	B15	B16	B17	B18
12.1	4.4	.0	2.8	1.2	2.3	3.7	4.6	.0	1.2	1.6	57.7	2.3	.2	7.5	10.5	.02	31	433
12.2	2.4	1.9	34.6	6.2	1.9	5.4	3.3	.0	2.9	1.9	24.0	1.4	.5	3.9	9.7	.01	27	208
12.3	3.8	1.0	4.8	1.0	1.9	3.4	5.3	.0	7.1	13.3	44.8	.0	.0	3.5	10.1	.01	28	210
12.4	2.3	.0	.0	.5	10.3	3.3	.9	.0	2.8	2.3	61.0	3.3	.0	3.3	10.0	.00	21	213
12.5	4.2	1.1	3.2	2.7	10.1	7.0	7.4	.0	2.7	3.2	42.0	1.1	1.1	4.2	9.4	.05	25	188
12.6	4.8	1.6	.0	.4	1.2	3.2	1.6	.0	4.8	3.2	64.1	6.5	.0	2.4	6.2	.00	21	248
12.7	3.8	2.4	3.8	.5	4.3	8.7	4.3	.5	1.0	1.0	38.6	5.7	1.0	12.2	14.2	.04	32	210
12.8	3.1	5.2	5.2	1.3	6.5	1.3	.4	.4	4.3	3.9	30.9	3.9	.9	7.3	25.4	.00	28	230
12.9	4.2	.5	.0	.5	4.6	4.7	.5	.0	4.1	1.8	46.8	8.3	2.8	7.9	13.3	.01	27	218
12.10	3.0	.0	.0	.4	4.6	6.8	2.1	.0	3.4	3.8	52.7	3.8	.4	7.1	11.9	.03	26	237
12.11	4.1	1.8	5.4	1.4	5.0	1.9	3.2	.5	2.3	3.2	46.8	8.3	1.8	6.6	10.1	.03	28	222
11.1	2.8	19.0	8.3	.8	5.6	5.2	.4	7.9	2.0	1.6	19.4	2.0	5.2	11.2	8.6	.76	27	252
11.2	2.7	17.9	5.0	.4	2.7	3.0	1.1	2.7	2.3	1.9	14.9	3.4	14.9	14.9	12.2	.66	27	262
11.3	4.2	26.3	2.5	.4	5.0	2.5	1.3	.8	2.9	1.3	15.0	2.1	16.3	13.4	6.0	.84	25	240
11.4	6.9	16.6	3.7	.4	2.4	2.8	2.0	7.3	4.1	.4	13.8	3.7	22.4	5.6	7.9	.24	28	246
11.5	3.4	13.8	3.8	.0	1.9	4.8	3.8	2.9	3.8	1.9	16.2	1.9	16.7	14.8	10.3	.34	30	210
11.6	2.5	19.0	5.0	.0	3.0	3.0	3.0	6.5	2.5	1.0	17.0	1.5	12.0	13.5	10.5	.93	23	200
11.7	1.1	18.6	3.8	.4	4.6	.3	2.3	4.2	1.9	.8	16.7	4.2	19.4	10.3	8.7	.40	26	263
11.8	3.6	16.4	2.5	.4	4.3	2.5	2.9	2.5	1.4	1.1	12.5	3.6	27.5	9.3	9.5	.44	26	280
11.9	3.7	6.7	5.2	1.5	5.2	3.7	4.4	8.9	3.7	.0	11.9	5.2	14.8	7.4	17.7	.11	21	135
11.10	2.2	31.1	1.9	.0	3.0	2.7	1.5	.7	3.4	.4	17.6	1.5	16.5	11.0	6.5	.75	26	267
11.11	1.9	16.7	2.7	.0	3.8	1.2	4.2	.8	3.0	.8	11.4	6.1	27.4	16.7	3.3	.29	24	263
11.12	1.7	10.9	1.1	.0	5.7	5.2	2.3	2.6	1.7	3	15.2	4.0	29.8	11.8	7.7	.16	29	375
11.13	.4	23.2	1.1	.0	6.4	9.7	2.2	2.5	4.6	.7	16.4	.4	16.1	10.5	5.8	.78	26	280
11.14	2.4	22.8	1.6	.0	5.3	3.2	2.0	.8	6.1	.0	13.4	1.2	18.3	17.8	5.1 <sup>1</sup>	.81	26	246
11.15	.7	21.5	1.7	.0	4.3	2.3	3.6	6.3	3.0	3	7.3	1.3	20.9	20.8	6.0	.80	25	302
11.16	.4	37.6	1.4	.0	2.2	3.6	2.5	9.0	3.2	.4	10.4	.7	11.5	12.3	4.8	.47	21	279
11.17	1.4	24.9	2.8	.5	5.1	3.7	4.6	8.8	3.2	.0	8.8	.5	8.3	20.9	6.5	.61	25	217
11.18	1.4	31.7	1.6	1.3	5.2	5.2	2.3	2.3	1.6	.0	11.8	2.6	15.4	13.7	3.9	.90	27	306
11.19	1.9	25.0	2.3	1.2	6.5	3.1	2.7	3.1	5.8	.4	13.5	1.5	15.4	10.7	7.2	.51	28	260

B1 = *Eoquittulina* spp.  
 B2 = *E. mosquensis*  
 B3 = *E. ubigi*  
 B4 = *Epistominina* spp.  
 B5 = *L. muensteri*  
 B6 = *Lenticulina* spp.  
 B7 = *Nodosaria/Dentalina* spp.  
 B8 = *P. rasanensis*  
 B9 = *S. elbrigata*  
 B10 = *S. infima*  
 B11 = *S. tenuissima*  
 B12 = *O. carinatum*  
 B13 = *O. strumosum*  
 B14 = Agglutinants  
 B15 = Restgroup  
 B16 = P/B-ratio  
 B17 = number of species  
 B18 = total number

Table 1. 3 Data set of new samples of Tojeira 1 section  
( F.M. Gradstein's data )

	1	2	3	4	5	6	7	8	9
T <sub>1</sub> /1	6.07	0.00	0.92	0.00	13.79	3.68	1.29	0.37	1.47
T <sub>1</sub> /2	1.97	0.00	2.96	0.00	20.20	6.90	1.97	0.00	0.49
T <sub>1</sub> /2A	1.99	0.28	1.42	0.28	23.01	7.95	3.41	0.85	0.85
T <sub>1</sub> /2B	5.68	0.44	0.00	0.00	18.34	1.75	2.62	0.00	1.31
T <sub>1</sub> /2C	2.49	0.83	5.81	0.00	7.88	7.05	3.73	0.00	2.07
T <sub>1</sub> /2D	1.80	0.36	0.72	0.00	19.42	3.24	1.44	0.00	3.24
T <sub>1</sub> /3	1.89	28.77	6.60	0.00	8.96	3.07	1.89	3.77	0.00
T <sub>1</sub> /4	4.40	11.00	3.77	0.00	12.26	1.89	10.06	7.86	15.09
T <sub>1</sub> /5	0.36	15.07	7.72	0.00	7.35	0.73	1.84	12.50	1.47
T <sub>1</sub> /6	0.49	27.36	4.97	0.00	13.93	1.24	0.99	14.42	4.98
T <sub>1</sub> /7	0.73	38.63	23.23	0.00	15.16	2.20	0.49	6.36	0.98
T <sub>1</sub> /8	0.59	25.09	4.72	0.00	24.10	2.36	2.94	9.43	3.24
T <sub>1</sub> /9	0.00	22.39	8.95	2.98	8.21	2.24	2.24	11.19	1.49
T <sub>1</sub> /10	0.39	18.82	15.29	0.39	11.76	1.57	2.74	7.84	4.70
T <sub>1</sub> /11	1.17	32.42	1.95	0.00	14.84	1.56	3.90	5.86	3.51
T <sub>1</sub> /12	0.99	25.74	1.98	0.00	7.42	1.73	2.97	3.46	2.47
T <sub>1</sub> /13	0.00	9.16	3.05	0.76	19.84	2.67	3.82	3.05	4.20
T <sub>1</sub> /14	0.00	18.94	3.28	0.00	18.43	2.27	4.80	10.10	3.03
T <sub>1</sub> /15	0.51	16.58	4.85	0.00	9.18	1.78	2.55	6.89	3.57
T <sub>1</sub> /16	0.00	39.05	1.90	0.00	25.71	2.40	0.95	1.90	2.40
T <sub>1</sub> /17	0.00	38.94	12.87	0.66	16.50	2.64	0.99	15.51	0.33

Table 1.3 (concluded)

	10	11	12	13	14	15	16	17	18
T <sub>1</sub> /1	9.19	30.83	0.73	1.47	25.56	4.60	0.00	26	544
T <sub>1</sub> /2	5.41	13.79	1.48	0.00	39.47	5.43	0.00	20	203
T <sub>1</sub> /2A	12.50	7.95	6.82	1.14	23.01	8.52	0.00	25	352
T <sub>1</sub> /2B	20.09	5.24	16.16	2.18	17.47	8.73	0.00	18	225
T <sub>1</sub> /2C	6.22	28.63	8.71	1.24	16.60	8.71	0.00	25	241
T <sub>1</sub> /2D	3.24	39.21	6.11	0.00	15.11	7.91	0.00	20	278
T <sub>1</sub> /3	11.38	17.69	3.07	0.00	10.61	2.36	0.03	22	424
T <sub>1</sub> /4	7.86	15.09	5.34	0.63	12.58	5.03	0.01	22	318
T <sub>1</sub> /5	0.00	18.75	18.01	4.04	9.93	2.20	0.01	18	272
T <sub>1</sub> /6	3.48	14.18	3.98	1.24	7.71	0.99	0.18	21	402
T <sub>1</sub> /7	0.49	2.20	2.69	2.93	2.44	1.47	0.18	22	409
T <sub>1</sub> /8	2.65	11.80	4.13	1.18	5.01	0.00	0.02	18	339
T <sub>1</sub> /9	1.49	6.72	0.00	0.00	32.09	0.00	0.08	15	134
T <sub>1</sub> /10	5.10	15.68	7.84	0.00	7.84	0.00	0.16	20	255
T <sub>1</sub> /11	1.95	9.37	4.69	5.86	10.55	2.34	0.31	24	256
T <sub>1</sub> /12	3.22	26.48	6.43	5.69	9.40	1.98	0.18	32	404
T <sub>1</sub> /13	0.00	19.08	4.96	10.30	17.94	1.14	0.06	21	262
T <sub>1</sub> /14	0.50	9.85	0.50	18.18	8.58	1.51	0.20	22	396
T <sub>1</sub> /15	2.29	17.86	26.02	0.26	4.33	3.32	0.08	22	392
T <sub>1</sub> /16	2.85	10.47	4.76	2.85	3.33	1.42	0.77	19	210
T <sub>1</sub> /17	0.00	5.28	0.66	1.32	3.30	0.99	0.78	22	303

Table 1.4 Data set of the new samples of Tojeira 2 section

( F.M. Gradstein's data ).

	1	2	3	4	5	6	7*	8	9 <sub>1</sub>
T <sub>2</sub> /1	0.40	0.00	69.70	17.82	2.37	0.00	0.40	0.00	1.78
T <sub>2</sub> /2	0.90	0.00	0.00	0.00	1.81	6.04	0.90	0.00	2.72
T <sub>2</sub> /3	0.59	0.00	3.55	1.77	2.96	11.24	1.77	0.00	2.96
T <sub>2</sub> /4	0.00	0.00	3.09	0.00	1.03	13.40	2.06	0.00	2.06
T <sub>2</sub> /5	0.35	1.05	28.67	28.67	1.40	15.73	1.08	0.00	0.35
T <sub>2</sub> /6	0.49	0.00	4.90	0.00	11.27	16.67	1.96	0.00	0.00
T <sub>2</sub> /7	0.00	47.59	11.44	13.85	4.22	3.67	0.60	1.80	0.60
T <sub>2</sub> /8	0.46	51.15	3.69	8.29	6.91	3.69	1.38	0.00	0.00
T <sub>2</sub> /9	0.00	25.22	10.62	6.64	1.33	3.10	0.00	5.31	2.65
T <sub>2</sub> /10	0.00	0.43	0.87	0.00	2.17	12.17	0.87	0.00	0.87
T <sub>2</sub> /11	0.00	23.7	5.18	7.41	16.3	0.00	1.48	0.00	1.48
T <sub>2</sub> /12	1.28	35.26	3.20	5.77	5.13	7.69	0.00	0.20	0.00
T <sub>2</sub> /13	0.00	54.06	1.63	0.00	9.35	2.44	2.03	3.66	2.44

1 = Eoguttulina sp.2 = Epistomina mosquensis3 = Epistomina uhligi4 = Epistomina sp.5 = Lenticulina muensteri6 = Lenticulina sp.7 = Nodoseria/Dentalina sp.8 = Pseudolamarckina rjasanensis9 = Spirillina elongata

Table 1.4. ( concluded )

	10	11	12	13	14	15	16	17	18
T <sub>2</sub> /1	1.78	1.19	0.00	0.00	1.39	1.58	0.00	20	520 <sub>2</sub>
T <sub>2</sub> /2	13.29	62.84	0.30	0.00	3.32	7.85	0.00	15	331
T <sub>2</sub> /3	5.92	37.28	0.59	0.59	14.20	16.57	0.00	19	169
T <sub>2</sub> /4	0.00	24.74	0.00	2.06	51.55	0.00	0.00	15	097
T <sub>2</sub> /5	1.05	3.15	0.00	0.00	6.99	11.54	0.00	19	286
T <sub>2</sub> /6	2.45	42.65	0.98	0.00	12.25	6.37	0.00	16	204
T <sub>2</sub> /7	1.20	3.01	1.20	0.00	3.01	7.83	0.48	17	166
T <sub>2</sub> /8	0.00	5.99	2.30	5.53	9.22	1.38	0.40	20	217
T <sub>2</sub> /9	0.00	12.83	0.44	2.65	25.67	3.54	0.23	20	226
T <sub>2</sub> /10	6.52	33.04	3.45	3.04	18.26	8.26	0.01	20	230
T <sub>2</sub> /11	7.41	11.85	1.48	3.70	12.6	7.41	1.21	26	135
T <sub>2</sub> /12	1.28	8.33	2.56	1.92	8.97	15.38	0.83	20	156
T <sub>2</sub> /13	0.41	9.76	1.63	1.21	8.94	2.44	1.06	20	246

10 = Spirillina infima

11 = Spirillina tenuissima

12 = Ophthalmidium carinatum

13 = Ophthalmidium strumosum

14 = agglutinants

15 = Restgroup

16 = P/B - ratio

17 = Number of species

18 = Total number

## CHAPTER 2: Statistical time series analysis

### 2.1 Introduction

Before applying the geostatistical modelling, time series models were applied as a possible strategy for explaining the available data structure from a statistical perspective.

Time series analysis does not have to fit the available data exactly. They can provide a model they only approximates the time process as long as the model explains the behavior of the available realization in a statistically adequate manner. Statistical tests are available to establish whether time series explain the data structure or not.

It is noted that most statistical calculations for this project were done using the Statistical Analysis System ( SAS ). For this reason, a brief overview of the SAS system is given before the actual statistical tests are described later.

### 2.2 SAS package programs

The SAS system is a software package for data analysis. The SAS language has its own vocabulary and syntax - words and the rules for putting them together.

The statements in a SAS program are divided into two kinds of steps; DATA steps and PROC steps ( or procedure steps ). These are the building blocks of all SAS programs. Usually DATA steps create SAS

data sets and PROC steps process SAS data sets: DATA and PROC steps can appear in any order and any number of data sets can be used. The DATA step begins with a data statement and can include any number of program statements. The PROC step asks SAS to call a procedure from its library, and execute that procedure. The PROC step begins with a PROC statement.

SAS output consists of two main parts:

- 1 - The SAS log which contains the SAS statements used in the job, notes, and error messages.
- 2 - The SAS procedure output ( File name , listing ).

During this research, several SAS software systems were used to assist in the geostatistical modelling.

SAS/Basics software is for the structure of the SAS language and its procedures are fundamental to the ( SAS ) system.

SAS/Statistics software is for the statistical analysis procedures in the SAS system. They range from simple descriptive statistics to complex multivariate techniques. Essentially, the user communicates with the SAS system by giving it a data file and the necessary PROC statements.

SAS/Graph software is a device - intelligent system for producing information and presenting it graphically. The software can be combined with other software in the SAS system. Version 5 SAS/Graph software runs on the main frame with the IBM 3179G, eight colour graphics terminal at the University of Ottawa.

SAS/ETS software is a package for econometrics and time series analysis. Procedures include;

- Time series extraction and plotting,
- Forecasting and time series analysis,
- Modeling

### 2.2.1 SAS package program ARIMA

The ARIMA model and techniques are described by Box and Jenkins (1976). This model is an algebraic statement showing how a time series variable ( $X_t$ ) is related to its own past values ( $X_{t-1}$ ,  $X_{t-2}$ ,  $X_{t-3}$ ). Consider the algebraic expression,

$$X_t = C + \phi_1 X_{t-1} + a_t \quad (1)$$

In Eq. 1,  $X_t$  is related to its own immediately past value ( $X_{t-1}$ ), where  $C$  is a constant term,  $\phi_1$  is a fixed coefficient whose value determines the relationship between  $X_t$  and  $X_{t-1}$ , and the  $a_t$  term is a probabilistic "shock" (random error, white noise) element. The generating mechanisms for two common ARIMA processes are written as follows:

$$X_t = C + \phi_1 X_{t-1} + a_t \quad (2)$$

$$X_t = C - \theta_1 a_{t-1} + a_t \quad (3)$$

The term  $\phi_1 X_{t-1}$  is called an *autoregressive* (abbreviated AR) process. A first order autoregressive process is abbreviated AR(1). In general, the label AR(p), means that the longest time lag attached to an

AR term in that process, is  $p$  time periods. The term  $\theta_1 a_{t-1}$  is called a *moving - average* ( abbreviated MA ) process. Again , a first order process is abbreviated MA(1). The label MA( $q$ ) means that the longest time lag attached to a MA term in that process is  $q$  time periods. The following additional processes can be considered:

$$X_t = C + \phi_1 X_{t-1} + \phi_2 X_{t-2} + a_t \quad (4)$$

$$X_t = C - \theta_1 a_{t-1} - \theta_2 a_{t-2} + a_t \quad (5)$$

$$X_t = C + \phi_1 X_{t-1} - \theta_1 a_{t-1} + a_t \quad (6)$$

Eq.4 is an AR(2) process because it contains only AR terms and the maximum time lag on the AR terms is two. Eq.5 is an MA(2) process since it has only MA terms, with a maximum time lag on the MA terms of two. Eq.6 is an example of a mixed process: which contains both AR and MA terms. It is an ARMA(1,1) process.

Let the AR order of a process be designated  $p$ , where  $p$  is some non - negative integer. Let  $q$ , also a non - negative integer, be the MA order of the process. The value  $d$ , another non - negative integer stands for the number of times a realization must be differenced to achieve a stationary mean. After a differenced series has been modeled, it is integrated  $d$  times. The letter " I " in the ARIMA refers to this integration step, and it corresponds to the number of times (  $d$  ) the original series has been differenced. ARIMA processes are characterized by the values of  $p$ ,  $d$ , and  $q$  in this manner;  
ARIMA (  $p$ ,  $d$ ,  $q$  ) ( AutoRegressive Integrated Moving Average ).

It is convenient to write ARIMA models in backshift notation using the multiplicative backshift operator B. The operator B is defined such that any variable which it multiplies has its time subscript shifted back by the power of B. This is shown in Eq.7.

$$B^k X_t = X_{t-k} \quad (7)$$

A constant is unaffected when multiplied by B, since, a constant has no time subscript.

$$B^k C = C \quad (8)$$

This model has two advantages over many other traditional single - series methods. The first advantage is that ARIMA is derived from a solid foundation of classical probability theory and mathematical statistics. Many other historically popular univariate methods are derived by intuitive methods. The second advantage is that ARIMA models are a family of models, not just a single model. Box and Jenkins ( 1964 ) have developed a strategy that guides the analyst in choosing one or more appropriate models out of this larger family of models. However, the construction of proper ARIMA models may require more experience and computer time than some of the other univariate methods.

The ARIMA method applies only to stationary time series. Although many non - stationary series arise in practice, most can be transformed into stationary series.

In ARIMA analysis, the observations in a single time series (one variable - univariate) are assumed to be statistically dependent, that is, sequentially or serially correlated.

The term  $X_t$  is a numerical value of an observation. The subscript  $t$  refers to the time period. A sequence of  $N$  observations can be represented as:  $X_1, X_2, X_3, \dots, X_N$ . The ARIMA model describes how any given observation ( $X_t$ ) is related to previous observations ( $X_{t-1}, X_{t-2}, \dots$ ). The mean of a stationary series indicates the overall level of this series. The sample mean is calculated as,

$$\bar{x} = \frac{1}{N} \sum_{t=1}^N X_t \quad t = 1, 2, \dots, N \quad (9)$$

The sample variance  $s^2(x)$ , is calculated as

$$s^2(x) = \frac{1}{N} \sum_{t=1}^N (X_t - \bar{x})^2 \quad (10)$$

When the mean of a series is stationary, a new series  $W_t$  is created. The series  $W_t$  is calculated as  $X_t$  minus sample mean. Where the sample mean ( $\bar{x}$ ) is calculated as before.

$$W_t = X_t - \bar{x} \quad (11)$$

The new series ( $W_t$ ) is indistinguishable from the series  $X_t$ , except

that it has a mean of zero ( $W_t = 0$ ). In fact, the two series  $X_t$  and  $W_t$  have all the same statistical properties except for their means. For example, the variances of the two series are identical.

### 2.2.2 Autocorrelation

The objective in autocorrelation analysis is to calculate a correlation coefficient ( $r_k$ ), for each set of ordered pairs ( $W_t, W_{t+k}$ ).

The  $k$  subscript represents the number of lags or time periods within the series. After calculating estimated autocorrelation coefficients, they are plotted graphically using an estimated autocorrelation function.

The autocorrelation coefficient ( $r_k$ ) measures the direction and strength of the statistical relationship between ordered pairs of observations on two random variables. It is a dimensionless number that can take on values only between  $-1$  and  $+1$ . A value of  $-1$  means perfect negative correlation and a value of  $+1$  means perfect positive correlation. If  $r_k = 0$  then  $X_{t+k}$  and  $X_t$  are not correlated at all in the available data. The formula for calculating the autocorrelation coefficients is,

$$r_k = \frac{\sum_{t=1}^{N-k} (X_t - \bar{x})(X_{t+k} - \bar{x})}{\sum_{t=1}^N (X_t - \bar{x})^2} \quad (12)$$

Eq.(12) can be written more compactly, because  $W_t$  is defined as  $X_t - \bar{x}$

Eq.(12) becomes,

$$r_k = \frac{\sum_{t=1}^N W_t W_{t+k}}{\sum_{t=1}^N (W_t)^2} \quad (13)$$

Many authors suggest that the maximum number of useful estimated autocorrelations is roughly  $N/4$  where  $N$  is the number of observations.

### 2.3 Partial autocorrelation

An estimated partial autocorrelation function is broadly similar to an estimated autocorrelation coefficient function (a.c.f.). The idea of partial autocorrelation analysis is to see how  $W_t$  and  $W_{t+k}$  are related. The estimated partial autocorrelation coefficient measuring this relationship between  $W_t$  and  $W_{t+k}$  is designated as  $\hat{\phi}_{kk}$ , which is an estimation of the correspondence parameters  $\phi_{kk}$ . The most accurate way of calculating partial autocorrelation coefficients (p.a.c.f.) is to estimate a series of least squares regression coefficients. Estimated partial autocorrelation coefficients are also found by applying regression techniques. The first step is to consider the true-regression relationship between  $W_{t+1}$  and the preceding value  $W_t$ . This is shown in Eq.(14).

$$W_{t+1} = \phi_{11} W_t + u_{t+1} \quad (14)$$

The term  $\phi_{11}$  is the partial autocorrelation coefficient to be estimated for  $k = 1$ ,  $u_{t+1}$ , which in turn, is the error term representing all the

factors affecting  $W_{t+1}$  that do not appear elsewhere in the regression equation. This is because there are no other  $W$ 's between  $W_k$  and  $W_{t+1}$ . However, there is a slightly less accurate, though easily computed, way to estimate the  $\phi_{kk}$  coefficients. It involves using the previously calculated autocorrelation coefficients ( $r_k$ ). As long as the data series is stationary, the following set of recursive equations gives fairly good estimates of the partial autocorrelations (Yule - Walker equations).

$$\hat{\phi}_{11} = r_1$$

$$\hat{\phi}_{kk} = \frac{\sum_{j=1}^{k-1} \hat{\phi}_{k-1,j} r_{k-j}}{1 - \sum_{j=1}^{k-1} \hat{\phi}_{k-1,j} r_j} \quad (k = 2, 3, \dots)$$

where  $\hat{\phi}_{kj} = \hat{\phi}_{k-1,j}$   $\left. \begin{matrix} \hat{\phi}_{kk} \\ \hat{\phi}_{k-1,k-j} \end{matrix} \right\} \quad (k = 3, 4, \dots, j = 1, 2, \dots, k-1)$

If  $X_t$  is a stationary normal distribution, then it has a constant expected mean  $\mu = E(X_t)$  for all  $X$ 's where  $E(\dots)$  denotes mathematical expectation.

$$\mu = E(X_t) = E(X_{t+m}) \quad (16)$$

The estimated constant variance,  $\sigma^2(x) = \nu_0(x) = E(X_t - \mu)^2$  for all  $X$ 's satisfies

$$\sigma^2(x) = E(X_{t+m} - \mu)^2 \quad (17)$$

and the constant autocovariance is

$$\nu_k(x) = E[(X_t - \mu)(X_{t+m} - \mu)]$$

For any two  $X$ 's separated by  $k$  time periods ( LAG ), the constant autovariance is,

$$\nu_k = E [ (X_k - \mu)(X_{t+k} - \mu) ] = E [ (X_{t+m} - \mu)(X_{t+k+m} - \mu) ]^2$$

When  $k = 0$ ,  $(X_t - \mu)$  times  $(X_{t+k} - \mu)$  is simply  $(X_t - \mu)^2$  in which case  $\nu_k(x) = 0 = \nu_0(x) = \sigma^2(x)$ . The estimated autocorrelation coefficients are derived as shown below

$$\rho_k = \frac{\nu_k}{\nu_0} \quad (18)$$

The estimation of auto and partial correlation coefficient values could be chosen according to different criterion. Box and Jenkins (1976) favor choosing according to the maximum likelihood ( ML ) criterion. However ML estimates of ARIMA procedure may require relatively large amounts of computer time. For this reason, Box and Jenkins (1976) suggest using the least - squares ( LS ) criterion. LS estimates, are either exactly or very nearly ML estimates.

#### 2.2.3.1 Testing autocorrelation coefficients

An estimated autocorrelation coefficient ( $r_k$ ) and theoretical autocorrelation coefficient ( $\rho_k$ ) are not exactly equal to each other due to sampling error. Bartlett (1946) has derived an approximate expression for the standard error of the sampling distribution of the  $r_k$  values. This estimated standard error  $s(r_k)$  is calculated as follows,

$$s(r_k) = (1 + 2 \sum r_j^2)^{1/2} N^{-1/2} \quad (19)$$

using the null hypothesis  $H_0 : \rho_k = 0$  for  $k = 1, 2, \dots, N$ . It is common when using the estimated  $s(r_k)$  in place of the true standard error  $s(r_k)$  to refer to the  $t$ -distribution in this way,

$$t_{r_k} = \frac{r_k - \rho_k}{s(r_k)} \quad (20)$$

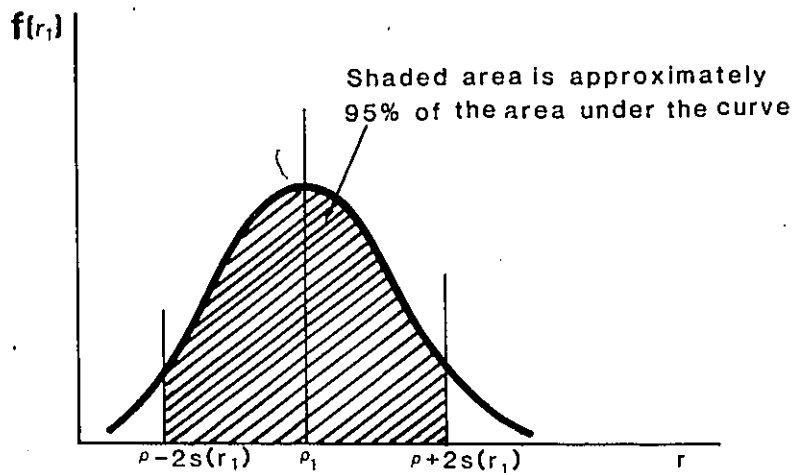
Fig. 2.1 illustrates these ideas. The label on the horizontal axis shows that this is a distribution of all possible values of  $r_1$  for a certain sample size  $N$ . This distribution is centered on the parameter  $\rho_1$  which is unknown. Since this is approximately a normal (or  $t$ ) distribution with an estimated standard error,  $s(r_1)$  given by Eq. 19, the interval  $\rho_1 \pm 2s(r_1)$  contains about 95% of all possible  $r_1$  values. This is represented by the shaded area under the curve. If MA is white noise, there would be no autocorrelation within the process and the MA order would be zero. In that case Eq. 19 would be

$$s(r_k) = N^{-1/2} \quad (21)$$

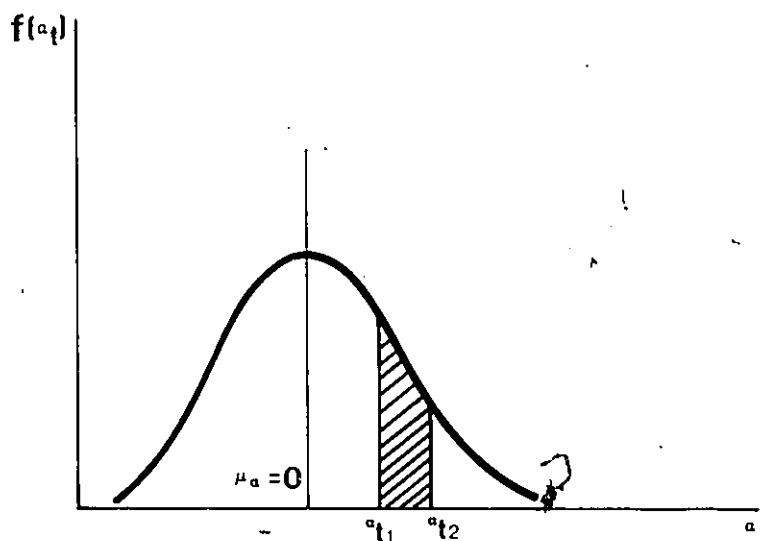
for all  $k$ .

### 2.2.3.2 Testing partial autocorrelation coefficients

Partial autocorrelation coefficients can also be tested and their statistical significance estimated. The required estimated standard error is



2.1



2.2

- Figure 2.1 Example of approximately normal ( or t ) sample distribution for  $r$  with estimated standard error  $s(r_1)$  given by Eq. 19
- Figure 2.2 Normal distribution of white noise ( random shock )  $a_t$  ( after Pankratz, 1983 )

$$s(\hat{\phi}_{kk}) = N^{-1/2} \quad (22)$$

Testing the null hypothesis  $H_0 : \phi_{11} = 0$  and  $t$ -statistic:

$$t_{\hat{\phi}_{11}} = \frac{\hat{\phi}_{11} - \phi_{11}}{s(\hat{\phi}_{11})} \quad (23)$$

### 2.2.3.3 Stationarity

Stationarity implies that the AR coefficients must satisfy certain conditions. If  $p = 0$ , the series are either a pure MA model or a white noise series. All pure MA models and white noise are stationary, so there are no stationary conditions to check. For an AR(1) or ARMA(1,q) process, the stationarity requirement is that the absolute value of  $\phi_1$  must be less than one ( $|\phi_1| < 1$ ). The  $\phi_1$  is not known, instead of it  $\hat{\phi}_1$  can be used. Table 2.1 shows a summary of stationary conditions for the ARMA procedure.

TABLE 2.1 Stationary conditions of time series (after Pankratz, 1983)

Model type	Stationarity conditions
ARMA ( 0,q )	Always stationary
AR( 1 ) or ARMA( 1,q )	$ \phi_1  < 1$
AR( 2 ) or ARMA( 2,q )	$ \phi_2  < 1$ $\phi_2 + \phi_1 < 1$ $\phi_2 - \phi_1 < 1$

### 2.3 The random shock ( white noise, random error ) model

The  $a_t$  terms in an ARIMA process are usually assumed to be normally indential and independently distributed random variables with a mean of zero and a constant variance. Those variables are often called " white noise ". Fig. 2.2 illustrates this idea with a normal distribution centered on zero. The horizontal axis shows the values are  $a_t$  independently distributed and therefore not autocorrelated. That is, knowing the set of past random shocks ( $a_{t-1}, a_{t-2}, a_{t-3}, \dots$ ), would not help to predict the current shock  $a_t$ . Random shocks ( $a_t$ ) are statistically independent, meaning that they are not autocorrelated. They can not be observed in practice but, they can be estimated. The residuals ( $\hat{a}_t$ ) are calculated from the estimated model. The residuals can be tested.

If the random shocks are serially correlated, then there is an autocorrelation pattern in  $X_t$  that has not been accounted for by the AR and MA terms in that model. If the residuals are autocorrelated, they are not white noise, so that the other model should be searched for residuals.

A residual autocorrelation coefficient is basically the same as any other estimated autocorrelation coefficient. The only difference is that the residuals ( $\hat{a}_t$ ) from an estimated model instead of observations in a realization ( $X_t$ ), are used to calculate the autocorrelation coefficients. Eq. 11 is used for this calculation,

$$r_k(\hat{a}) = \frac{\sum_{t=1}^{N-k} (\hat{a}_t - \bar{a})(\hat{a}_{t+k} - \bar{a})}{\sum_{t=1}^N (\hat{a}_t - \bar{a})^2} \quad (24)$$

where  $r_k(\hat{a})$  is the residual autocorrelation coefficient.

If the random shocks are uncorrelated, estimation of  $(\hat{a})$  should also be uncorrelated. Therefore, the residual autocorrelation coefficients for an ARIMA model will ideally have autocorrelation coefficients that are all statistically zero. But all residual autocorrelations cannot be expected to be zero. The reason is, that the residuals are calculated from a calculation using only estimates of the ARIMA coefficients (not their true values). Therefore, the sampling error will cause some residual autocorrelations to be nonzero.

The AR(1) procedure can be written as  $(1 - \phi B)W_t = a_t$  or

$$X_t = \mu(1 - \phi_1) + \phi_1 X_{t-1} + a_t \quad (25)$$

The random shock  $a_t$  can not be observed during the period  $t - 1$  but  $\mu$ ,  $\phi_1$  and  $X_{t-1}$  are known parameters. Using  $\mu$ ,  $\phi_1$  and  $X_{t-1}$  to find the calculated value of  $X_t$  designated  $\hat{X}_t$ :

$$\hat{X}_t = \mu(1 - \phi_1) + \phi_1 X_{t-1} \quad (26)$$

at time  $t$ ,  $X_t$  can be observed, then the random shock  $a_t$  can be found by

subtracting the calculated value  $\hat{X}_t$  from the observed value  $X_t$  in Eq 25

$$X_t - \hat{X}_t = a_t \quad (27)$$

$\mu$  and  $\phi_1$  are known, but in practice, parameters of the ARIMA procedure are not known. They must be estimated from the data. These estimates in the present case, are  $\hat{\mu}$  and  $\hat{\phi}_1$ . Modifying ( 26 ) accordingly, the calculated value  $\hat{X}_t$  becomes

$$\hat{X}_t = \hat{\mu} ( 1 - \hat{\phi}_1 ) + \hat{\phi}_1 X_{t-1} \quad (28)$$

When  $\hat{X}_t$  is calculated from estimates of parameters rather than known parameters, ( 27 ) does not give the exact value of the random shock  $a_t$ . An estimate of the random shock  $a_t$  is denoted  $\hat{a}_t$  and called a residual:

$$X_t - \hat{X}_t = \hat{a}_t \quad (29)$$

Eq. 29 is the definition of a residual for any ARIMA model.

AR(1),  $\hat{X}_t$  depends on  $\hat{\mu}$  and  $\hat{\phi}_1$  along with  $X_{t-1}$ , as shown Eq. 28

MA(1),  $\hat{X}_t$  depends on  $\hat{\mu}$  and  $\hat{\theta}_1$  along with  $\hat{a}_{t-1}$ .

$$\hat{X}_t = \hat{\mu} - \hat{\theta}_1 \hat{a}_{t-1} \quad (30)$$

ARIMA( 1,1),  $\hat{X}_t$  depends on  $\hat{\mu}$ ,  $\hat{\phi}_1$ ,  $\hat{\theta}_1$  along with  $X_{t-1}$  and  $\hat{a}_{t-1}$ ;

$$\hat{X}_t = \hat{\mu} ( 1 - \hat{\phi}_1 ) + \hat{\phi}_1 X_{t-1} - \hat{\theta}_1 \hat{a}_{t-1} \quad (31)$$

$$\hat{C} = \hat{\mu} (1 - \hat{\phi}_1) \quad (32)$$

### 2.3.1 $t_b$ - test

Approximate  $t$  values are calculated for residual autocorrelation coefficients using Bartlett's approximation for standard error of estimated autocorrelations. The null hypothesis;  $H_0: \rho_k(\hat{a}) = 0$  can be tested for each residual autocorrelation coefficient and the number of standard errors ( $t$ ) away from zero each residual autocorrelation coefficient falls can be calculated.

$$t_b = \frac{r_k(\hat{a}) - 0}{s[r_k(\hat{a})]} \quad (33)$$

In practice, if the absolute value of a residual autocorrelation coefficient  $t$  - value is less than 1.25 at lags 1, 2, and 3, and less than about 1.6 at larger lags, it means that the random shocks at that lag are independent. If any residual autocorrelation coefficient  $t$  - value is larger than the critical values suggested above. The null hypothesis, is rejected and it is assumed that the random shocks from the estimated model are correlated.

### 2.3.2 Chi - squared test

There is another way of estimating residual autocorrelation coefficient  $t$  - values. Ljung and Box ( 1978 ) suggest a test statistic

based on all the residual autocorrelations as a set.  $K$  is a number of residual autocorrelations. The null hypothesis is tested about the correlations among the random shocks,

$$H_0: \rho_1(a_1) = \rho_2(a) = \dots = \rho_K(a) = 0$$

with this test statistic

$$Q = N(N+2) \sum_{k=1}^K (N-k)^{-1} r_k^2(\hat{a}) \quad (34)$$

where  $N$  is the number of observations using the estimated model. The statistic  $Q$  approximately follows a chi-squared distribution with  $(K - m)$  degrees of freedom, where  $m$  is the number of parameters estimated in the ARIMA model. If  $Q$  is large (significantly different from zero) then the residual autocorrelations, as a set, are significantly different from zero, and the random shocks of the estimated model are probably autocorrelated.

## 2.4 Periodogram

A time series with  $N$  observations can be represented by a trigonometric polynomial,

$$X_t = a_0/2 + \sum (a_k \cos w_k t + a_b \sin w_k t) \quad (35)$$

where  $\omega = \frac{2\pi k}{N}$   $k = 0, 1, 2, \dots, m$

$$a_k = \frac{\sum_{t=1}^N x_t \cos \omega_k t}{N} \quad (36)$$

$$b_k = \frac{\sum_{t=1}^N x_t \sin \omega_k t}{N} \quad (37)$$

$$a_0 = x/2$$

$m = N/2$  ( if  $N$  is even ) and  $m = (N - 1)/2$  ( if  $N$  is odd )

The periodogram is defined as

$$I_k(\omega_k) = N ( a_k^2 + b_k^2 ) / 2 \quad (38)$$

for the decomposition of the process into two - degrees - of - freedom components for each of the  $m$  frequencies. When  $N$  is even, the last periodogram value is a one - degree - of - freedom component.

If  $(X_t)$  is a sequence of normal independent  $(\mu, \sigma^2)$  random variables, then the  $a_k$  and  $b_k$ , being linear combinations of the  $X_t$ , will be normally distributed. Since the sine and cosine functions are orthogonal, ( Fuller, 1969 ), the  $a_k$  and  $b_k$  are independent.

Fuller ( 1969 ) suggested that the distributional properties of the periodogram ordinates are easily obtained when the time series is

normal white noise.

#### 2.4.1 Fisher's kappa test

A statistic that can be used to test the white noise hypothesis is

$$\xi = \left[ (1/m) \sum I_n(\omega_k) \right]^{-1} I_n(L) \quad (39)$$

where  $I_n(L)$  is the largest periodogram ordinate in a sample of  $m$  periodogram ordinates each with two degrees of freedom. A table of the distribution of  $\xi$  is given by Davis (1941).

Under the null hypothesis that a time series is normal white noise, the periodogram ordinates are multiples of independent chi-squares, each with two degrees of freedom.

If the largest ordinate is  $P_{\max}$ , the sum of all periodogram values is  $P$  and sample size is  $m - 1$ . Then Fisher's kappa test has

$$\text{Test Statistic} = (m - 1) P_{\max} / P \quad (40)$$

#### 2.4.2 Kolmogorov - Smirnov test

Bartlett (1946) suggested a test based on the normalized cumulative periodogram which is a uniform (0, 1) distribution. Therefore, the normalized periodogram as a sample distribution function is plotted and the Kolmogorov - Smirnov test of the hypothesis

function is plotted and the Kolmogorov - Smirnov test of the hypothesis is applied to tell whether the original time series is white noise or not.

The cumulative periodogram is plotted against  $k$ . The upper and lower bounds are drawn according to the Kolmogorov - Smirnov test. If the normalized cumulative periodogram passes the 5% or 95% boundary lines, the data rejects the hypothesis of independence.

Birnbaum ( 1952 ) indicated that for  $( m - 1 ) > 30$  , the 95% point for the Kolmogorov - Smirnov statistic is approximately  $1.36 ( m - 1 )^{1/2}$  and the 99% point is approximately  $1.63 ( m - 1 )^{1/2}$ .

The periodogram was calculated with a SAS package program called the SPECTRA procedure. SPECTRA creates an output whose variables can contain values of the periodograms and estimates of spectral densities. The periodogram ordinates can be smoothed by a moving average to produce estimated spectral densities. SPECTRA can also test whether or not the data are white noise, according to Fisher's kappa and Bartlett's Kolmogorov - Smirnov test.

## CHAPTER 3: Geostatistical model:

### 3.1 Correlogram

A set of autocorrelation coefficients is a graph called a correlogram in which  $r_k$  is plotted against the lag  $k$ . The correlogram is helpful in identifying that type of ARIMA model gives the best representation of observed values.

For stationary series, the correlogram is compared with the theoretical autocorrelation functions of different ARMA processes in order to choose the one that is most appropriate. The autocorrelation coefficient function of the MA( $q$ ) process is easy to recognize as it will 'cut off' at lag  $q$ , whereas the autocorrelation coefficient function (a.c.f) of an AR( $p$ ) process is a mixture of damped exponentials and sinusoids and dies out slowly.

If a non-stationary series contains a trend, then the values of  $r_k$  may not come down to zero quickly. This indicates a non-stationary process.

### 3.2 Trend

Agterberg (1974a) suggested that the systematic variations called "trends" or the spatial variability of measurable features can be divided into regional variations of a systematic nature and more local, unpredictable, fluctuations.

AR processes have theoretical a.c.f.'s which tail off toward zero. This tailing off might follow a simple exponential decay pattern, a damped sine wave, or a more complicated decay or wave pattern. Fig. 3.1 shows the theoretical a.c.f.s and p.a.c.f.s for two types of stationary AR(1) processes. Any stationary AR(1) process has a theoretical a.c.f. showing exponential decay and a p.a.c.f. with an asterisk at lag 1. If  $\phi_1$  is positive the a.c.f. decays on the positive side and the p.a.c.f. asterisk is positive. This is illustrated by Fig. 3.1.

If  $\phi_1$  is negative, the AR(1) a.c.f. decays with alternating signs, starting from the negative side while the p.a.c.f. asterisk is negative. This is illustrated by Fig. 3.2.

MA processes have a theoretical p.a.c.f. which tails off to zero after lag  $q$ . This tailing off may be either some kind of exponential decay or some type of damped wave pattern. Fig. 3.3 shows two MA(1) theoretical a.c.f.'s and p.a.c.f.'s. Any MA(1) process has a theoretical a.c.f. with an asterisk at lag 1 followed by a cut-off to zero, and a theoretical p.a.c.f. which tails off toward zero.

If  $\theta_1$  is negative, the asterisk in the a.c.f. is positive whereas the p.a.c.f. decays exponentially, with an alternating sign starting on the positive side. This is illustrated by Fig. 3.3.

If  $\theta_1$  is positive, the a.c.f. asterisk is negative, while the p.a.c.f. decays exponentially on the negative side. This is illustrated by Fig. 3.4. ARMA processes or mixed processes have theoretical a.c.f.'s with both AR and MA characteristics. The a.c.f. tails off toward zero after the first  $q - p$  lags with either exponential decay or a damped

Figure 3.1 Examples of theoretical a.c.f.'s and p.a.c.f.'s for stationary AR(1) processes with  $\phi_1 > 0$

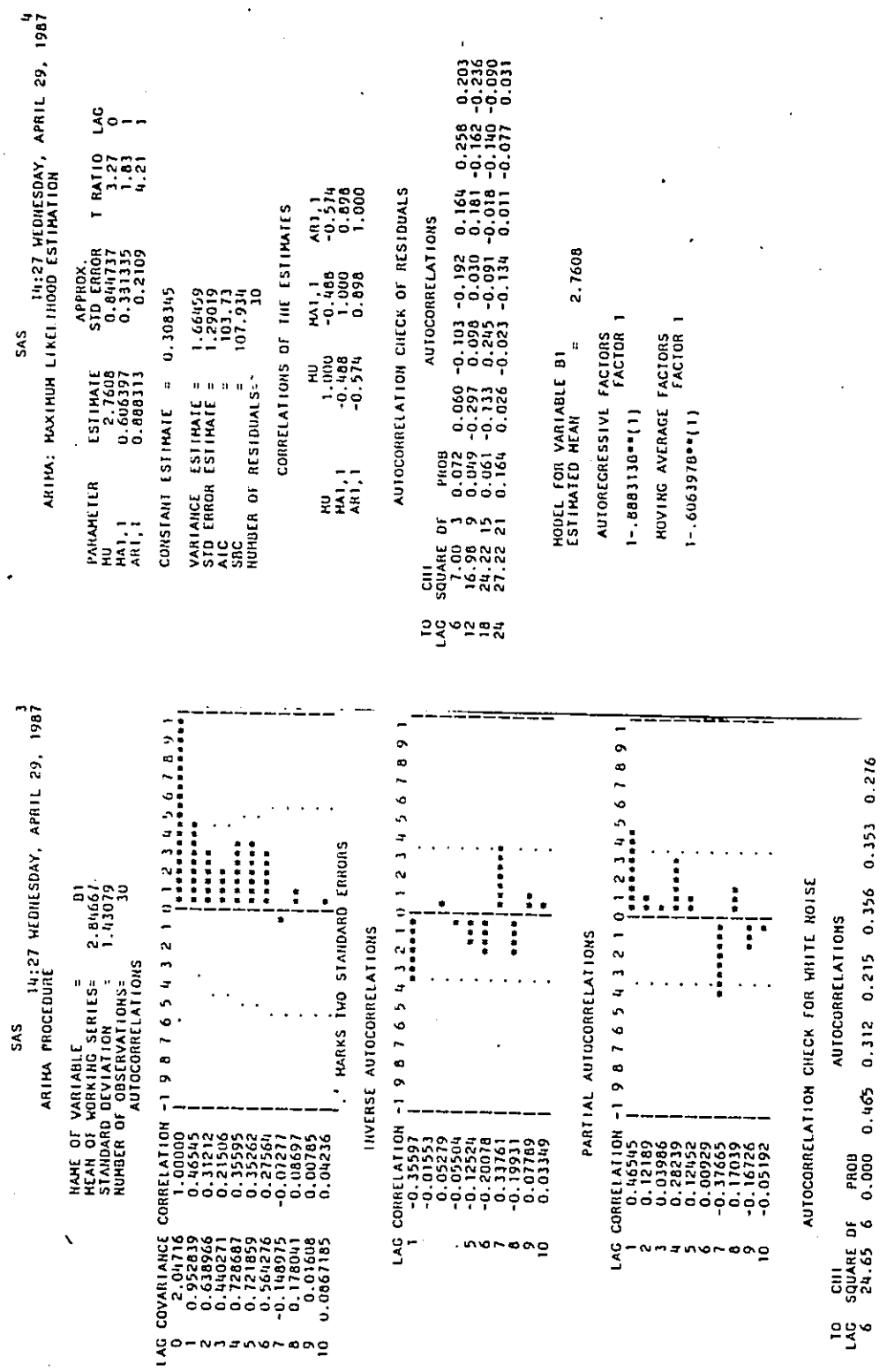


Figure 3.2 Examples of theoretical a.c.f.'s and p.a.c.f.'s for stationary AR(1) processes with  $\phi_1 < 0$

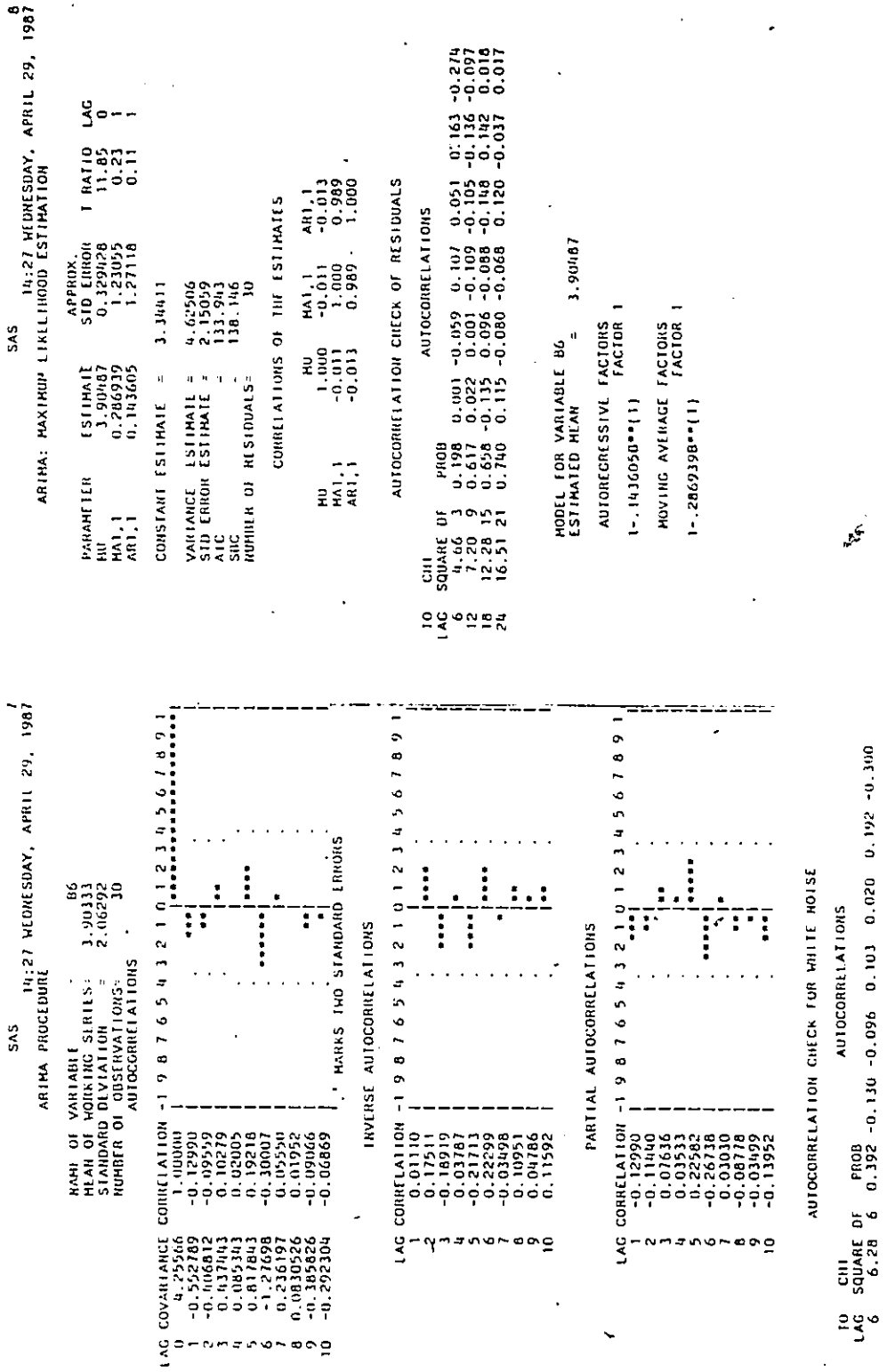


Figure 3.3 Examples of theoretical a.c.f.'s and p.a.c.f.'s for stationary AM(1) processes with  $\theta_1 < 0$

SAS 14:27 WEDNESDAY, APRIL 29, 1987  
 ARIMA: MAXIMUM LIKELIHOOD ESTIMATION

SAS 14:27 WEDNESDAY, APRIL 29, 1987  
 ARIMA PROCEDURE

NAME OF VARIABLE = A7  
 MEAN OF WORKING SERIES = 4.72516  
 STANDARD DEVIATION = 2.81477  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

PARAMETER ESTIMATE APPROX.  
 MU 4.72516 STD ERROR 8.275  
 MA1,1 -0.163017 0.187076 7.90 0  
 -0.87 1

CONSTANT ESTIMATE = 4.72516  
 VARIANCE ESTIMATE = 8.275  
 STD ERROR ESTIMATE = 2.87663  
 AIC 155.444  
 SBC 158.312  
 NUMBER OF RESIDUALS = 31

CORRELATIONS OF THE ESTIMATES  
 MU MA1,1  
 MU 1.000 0.030  
 MA1,1 0.030 1.000

LAG COVARIANCE CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1  
 0 7.92293 1.00000  
 1 1.10392 0.11933  
 2 -0.0278066 -0.00318  
 3 -1.22146 -0.15482  
 4 -1.00178 -0.13780  
 5 0.754833 0.09257  
 6 -0.727153 -0.09718  
 7 -0.841242 -0.10618  
 8 -0.607848 -0.07672  
 9 -1.96482 -0.24799  
 10 -1.67154 -0.21097

MARKS TWO STANDARD ERRORS

AUTOCORRELATION CHECK OF RESIDUALS

LAG	SQUARE	DF	PROB	AUTOCORRELATIONS
6	3.64	4	0.457	-0.006 -0.025 0.177 -0.185 0.14 -0.093
12	10.47	10	0.400	-0.082 -0.039 -0.221 -0.156 -0.219 0.105
18	16.63	16	0.410	-0.126 -0.119 0.013 -0.146 0.196 0.011
24	22.68	22	0.420	0.044 0.238 0.020 0.043 -0.050 -0.019

MODEL FOR VARIABLE AT  
 ESTIMATED MEAN = 4.72516  
 MOVING AVERAGE FACTORS  
 1+0.163017B\*\*(1)

INVERSE AUTOCORRELATIONS

LAG	CORRELATION
1	-0.38613
2	0.40548
3	-0.37965
4	0.37481
5	-0.30921
6	0.25903
7	-0.14434
8	0.15599
9	0.00262
10	0.16411

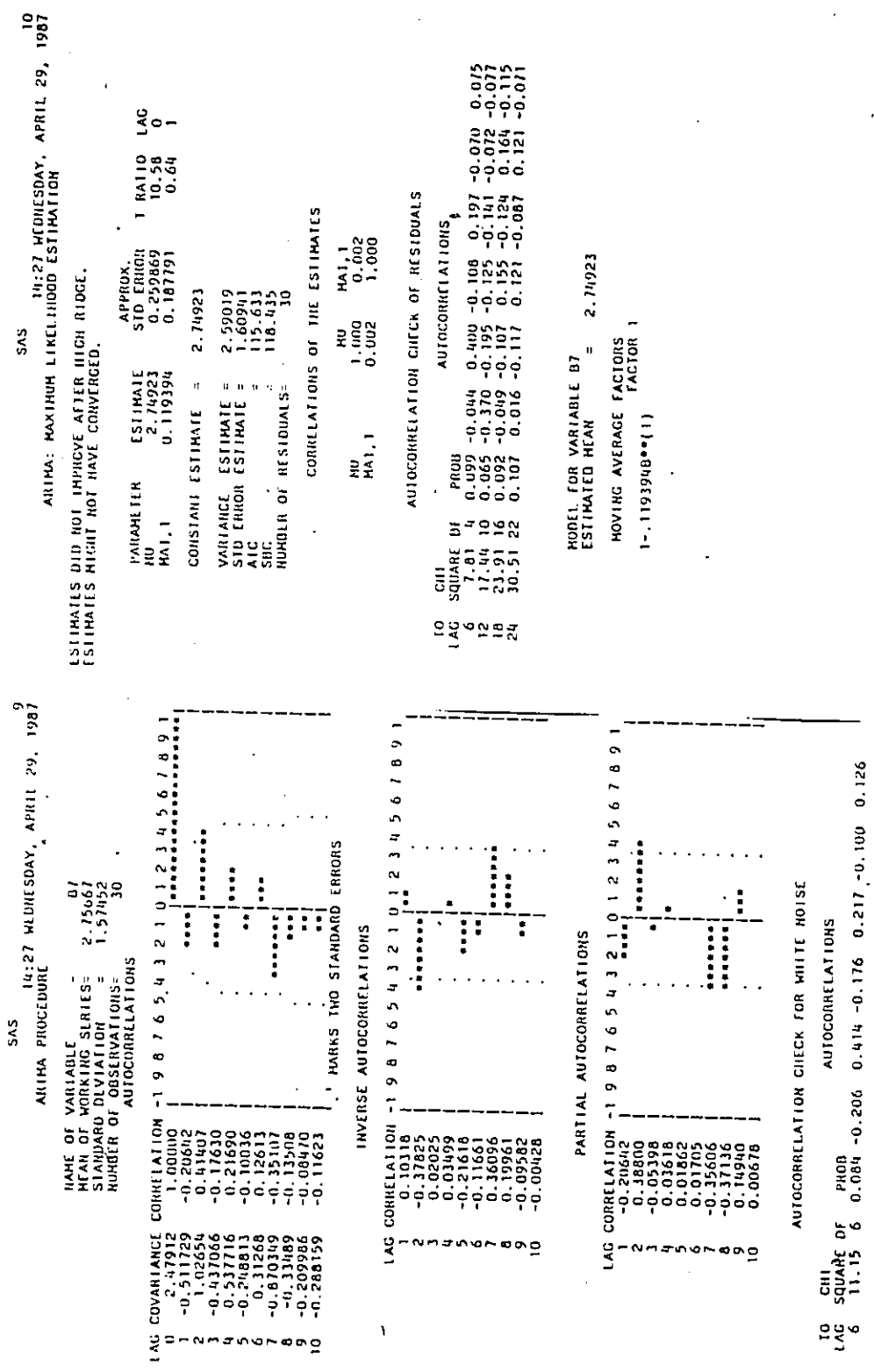
PARTIAL AUTOCORRELATIONS

LAG	CORRELATION
1	0.13933
2	-0.02335
3	0.16139
4	-0.19177
5	0.16925
6	-0.20026
7	0.03173
8	-0.18538
9	-0.11470
10	-0.25130

AUTOCORRELATION CHECK FOR WHITE NOISE

LAG	SQUARE	DF	PROB	AUTOCORRELATIONS
6	2.95	6	0.814	0.139 -0.003 0.154 -0.138 0.095 -0.092

Figure 3.4 Examples of theoretical a.c.f.'s and p.a.c.f.'s for stationary AM(1) processes with  $\theta_1 > 0$



7

sine wave. The theoretical p.a.c.f. tails off to zero after the first  $p - q$  lags. Fig. 3.5 shows theoretical a.c.f.'s and p.a.c.f.'s for some types of ARMA ( 1, 1 ) processes.

Non - stationary series show a trend, which may be either in the mean , the variance or both. The series can be decomposed into two parts. The first one, representing the mean of the series, accounts for the nonstationary trend by a deterministic function, which depends on the origin. The second one is a stochastic part with zero mean. Pandit et al. 1983 has shown that many of the nonstationary data can be modelled by explicitly including polynomial, exponential or sinusoidal functions, dependent on the choice of origin, to represent the mean of the series. Furthermore, the series with a nonstationary variance may be stationary in the natural logarithm.

### 3.2.1 Exponential trend

Exponential trends very often arise in the data obtained from scientific applications. Many of these systems can be well represented by differential equations with constant coefficients, and the nature of noise is generally unknown. If the constant of proportionality is denoted by  $c$ , then the equation of the system is

$$\frac{dX(t)}{dt} = - cX(t) \quad (41)$$

Figure 3.5 Examples of theoretical a.c.f.'s and p.a.c.f.'s two stationary ARMA(1,1) processes

SAS 14:27 WEDNESDAY, APRIL 29, 1987  
 ARIMA PROCEDURE

SAS 14:27 WEDNESDAY, APRIL 29, 1987  
 ARIMA PROCEDURE

PARAMETER ESTIMATE APPROX  
 STD. ERROR T RATIO LAG  
 MU 4.50075 0.483008 9.30 0  
 MA1,1 -0.647902 0.378028 -1.72 1  
 AR1,1 -0.298897 0.448744 -0.67 1

CONSTANT ESTIMATE = 5.84601  
 VARIANCE ESTIMATE = 4.44094  
 STD ERROR ESTIMATE = 2.10735  
 AIC = 132.909  
 SBC = 137.112  
 NUMBER OF RESIDUALS = 30

LAG COVARIANCE CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1  
 0 4.52766 1.00000  
 1 0.709677 0.15674  
 2 -1.67178 -0.35924  
 3 -0.614223 -0.13566  
 4 0.292387 0.06458  
 5 0.139454 0.03080  
 6 0.213731 0.04721  
 7 -0.0601248 -0.01328  
 8 -0.643481 -0.14212  
 9 -0.522414 -0.11538  
 10 0.0617407 0.01364

MARKS TWO STANDARD ERRORS

CORRELATIONS OF THE ESTIMATES  
 MU MA1,1 AR1,1  
 MA1,1 1.000 -0.003 1.000 0.904  
 AR1,1 0.009 0.904 1.000 1.000

AUTOCORRELATION CHECK OF RESIDUALS

LAG CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1  
 1 -0.21865  
 2 0.36645  
 3 0.02426  
 4 -0.02148  
 5 0.06662  
 6 0.01196  
 7 0.05845  
 8 0.10023  
 9 0.02535  
 10 0.04799

INVERSE AUTOCORRELATIONS

TO CHI LAG SQUARE DF PROB  
 6 6.42 6 0.378 0.157 -0.369 -0.136 0.065 0.031 0.047

AUTOCORRELATIONS

MODEL FOR VARIABLE B5 = 4.50075  
 ESTIMATED MEAN = 4.50075  
 AUTOREGRESSIVE FACTORS  
 1+0.298897B\*\*(1)  
 MOVING AVERAGE FACTORS  
 1+0.647902B\*\*(1)

LAG CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1  
 1 0.15674  
 2 -0.40372  
 3 0.01085  
 4 -0.06937  
 5 -0.02711  
 6 0.06307  
 7 -0.05135  
 8 -0.11133  
 9 -0.09157  
 10 -0.06065

PARTIAL AUTOCORRELATIONS

AUTOCORRELATION CHECK FOR WHITE NOISE

TO CHI LAG SQUARE DF PROB  
 6 6.42 6 0.378 0.157 -0.369 -0.136 0.065 0.031 0.047

AUTOCORRELATIONS

LAG SQUARE DF PROB  
 6 6.42 6 0.378 0.157 -0.369 -0.136 0.065 0.031 0.047

AUTOCORRELATIONS

SAS 21:31 TUESDAY, JULY 14, 1987  
 ARIHA: MAXIMUM LIKELIHOOD ESTIMATION

PARAMETER	ESTIMATE	APPROX STD ERROR	T RATIO	LAG
MU	1.0000	0.0000	2.72	0
MA1,1	-0.27989	0.12977	-2.17	1
AR1,1	0.254803	0.365735	0.70	1

CONSTANT ESTIMATE = 2.0183

VARIANCE ESTIMATE = 16.8526  
 STD ERROR ESTIMATE = 4.10319  
 SSC = 178.864  
 SSC = 182.966  
 NUMBER OF RESIDUALS = 31

CORRELATIONS OF THE ESTIMATES

MU	MA1,1	AR1,1
MU	1.0000	-0.022
MA1,1	-0.022	1.0000
AR1,1	-0.035	0.867

AUTOCORRELATION CHECK OF RESIDUALS

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	3.22	3	0.358	-0.009
12	3.36	9	0.948	0.010
18	4.58	15	0.995	-0.041
24	6.75	21	0.999	-0.075

MODEL FOR VARIABLE A2  
 ESTIMATED MEAN = 2.70841

AUTOREGRESSIVE FACTORS  
 1-.254803B\*\*(1)  
 MOVING AVERAGE FACTORS  
 1+0.278989B\*\*(1)

SAS 21:31 TUESDAY, JULY 14, 1987  
 ARIHA PROCEDURE

NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES = 2.77097  
 STANDARD DEVIATION = 4.48713  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG COVARIANCE CORRELATION

LAG	COVARIANCE	CORRELATION
0	20.1343	1.00000
1	9.83615	0.48853
2	4.60299	0.22861
3	5.48433	0.27205
4	5.80549	0.28834
5	1.74721	0.08678
6	-0.01352	-0.00897
7	-0.10928	-0.05272
8	-0.37689	-0.02168
9	-0.55789	-0.04846
10	-0.975797	

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG CORRELATION

LAG	CORRELATION
1	-0.39887
2	-0.15046
3	-0.17302
4	-0.12338
5	0.05685
6	0.03231
7	0.02655
8	0.03444
9	-0.04236
10	-0.01083

PARTIAL AUTOCORRELATIONS

LAG CORRELATION

LAG	CORRELATION
1	0.48853
2	-0.01352
3	0.28834
4	0.12338
5	-0.07326
6	-0.06374
7	-0.00149
8	0.04306
9	-0.01458
10	

AVG. CORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	17.22	6	0.009	0.489
			0.229	0.322
			0.489	0.288
			0.087	-0.001

or

$$\frac{dX(t)}{dt} + c X(t) = 0 \quad (42)$$

The solution of an  $n^{\text{th}}$  order linear differential equation has the form of a sum of  $n$  exponentials. The homogenous differential equation is,

$$\frac{df}{dt} = f \quad \text{or} \quad \frac{df}{dt} - f = 0 \quad (43)$$

To solve this, we want a function that gives back the same function after differentiation. Other than the trivial solution which is  $f(t) = 0$ , the simplest function that can be tried is  $f(t) = 1$ . If  $f(t)$  is differentiated, we must get 1 whereas  $f(t) = 1$  gives 0. In order to obtain 1, we must add  $t$  to the function. Let us try:

$$f(t) = 1 + t \quad (43a)$$

If 43a is differentiated we must get  $1 + t$ . To do this we would have to add  $t^2/2$  to the function. Thus,

$$f(t) = 1 + t + t^2/2 \quad (43b)$$

It is clear that the desired function is the infinite sum

$$1 + t + t^2/2! + t^3/3! + \dots + t^n/n! \text{ with } n \rightarrow \infty$$

where

$$n! = n (n - 1) (n - 2) \dots 3.2.1$$

This is the exponential function

$$e^t = \sum_{n=0}^{\infty} \frac{t^n}{n!} \quad (44)$$

which is a solution of  $\frac{df}{dt} - f = 0$ . This being a homogenous

differential equation, the general solution is,

$f(t) = c e^t$ . By similar reasoning, the solution of  $\frac{df}{dt} = \beta f$  becomes

$$e^{\beta t} = 1 + \beta t + \frac{(\beta t)^2}{2} + \dots + \frac{(\beta t)^n}{n!} + \dots \quad (45)$$

The general solution is  $f(t) = c e^{\beta t}$ . Integration of Eq. 42 gives

$$(\beta + a) c e^{\beta t} = 0 \quad (46)$$

This gives the characteristic equation

$\beta + a = 0$  and  $\beta = -a$ . The solution of Eq. 42 takes the form

$f(t) = c e^{-at}$  or for any time period (h) the correlogram shows

$$r(h) = c e^{-ah} \quad (47)$$

When a logarithmic scale is used for vertical axis, the exponential curve becomes a straight line.

### 3.3 Probit transformation

$S_0$  - called tolerance values in response to a stimulus vary from one sample to another in the population. If tolerance is measured

by  $\lambda$ , Finney ( 1952 ) suggested that the distribution of tolerances may be expressed as,

$$dP = f ( \lambda ) d\lambda \quad (48)$$

If a  $\lambda_0$  is assigned to the whole population, all individuals will respond whose tolerances are less than  $\lambda_0$ , and their proportion  $P$  in the population satisfies

$$P = \int_0^{\lambda_0} f ( \lambda ) d\lambda \quad (49)$$

If this equation is extended from 0 to  $\infty$ , it becomes,

$$\int_0^{\infty} f ( \lambda ) d\lambda = 1 \quad (50)$$

Usually, the distribution of the tolerance can be made approximately normal by an appropriate transformation.

Bliss ( 1934 ) first proposed the name " probit " for his modification of the normal equivalent deviate, which he increased by 5 so as to simplify the arithmetical procedure by avoiding negative values.  $P$  corresponds to a probability in a normal distribution with mean equal to 1. The probit of  $P$  is  $Y - 5$  and a variance equal to 1.

$$P = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{Y-5} e^{(-1/2)u^2} du \quad (51)$$

A table giving probits for specified values of P was prepared by Bliss (1935a). It was reproduced by Fisher and Yates (1963) as table IX of their Statistical tables for Biological, Agricultural and Medical Research. The values shown in their table IX are equal to fractiles ( $u$ ) of the normal distribution in standard form augmented by 5.

For example:  $u_{0.98} + 5 = 7.054$  and  $u_{0.98} = 2.054$

The effect of a transformation from sample percentages to probits is illustrated in Fig. 3.6. The normal curve becomes a straight line when the probit transformed values are plotted on probability paper.

### 3.4 Frequency Distributions

The average of  $N$  observations is defined as

$$y = \bar{x}_N = (1/N) (x_1 + x_2 + \dots + x_N) \quad (52)$$

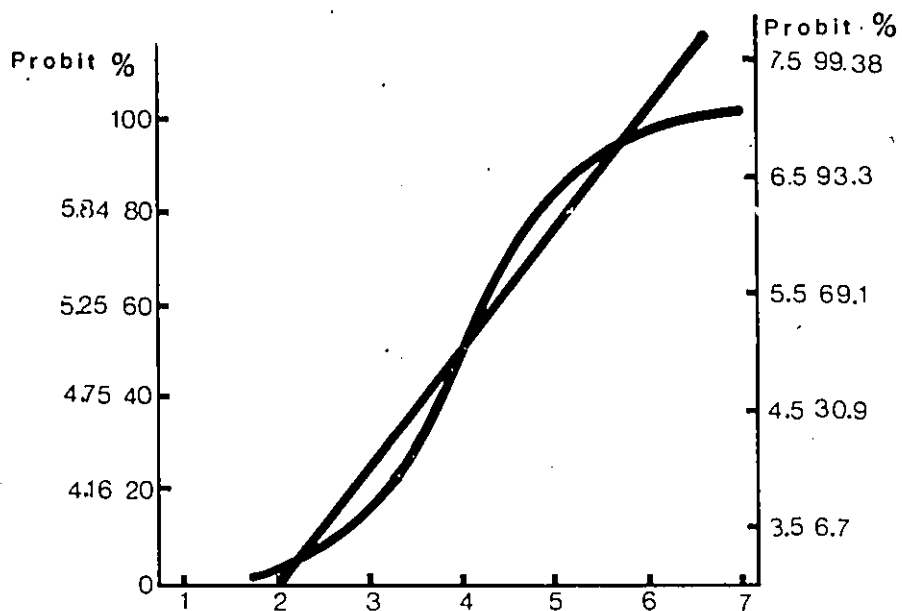
Suppose that the standard error is  $s(x)$ , the variance can be written as

$$s^2(y) = (1/N) s^2(x) \quad (53)$$

Eq. 52 represents one of the more important theorems of statistical theory. Eq. 53 can also be written as,

$$F_N = N s^2(\bar{x}) / s^2(x) = 1 \quad \text{or} \quad F_N = \frac{N s^2(x_N)}{s^2(x)} = 1 \quad (54)$$

The factor  $F_N$  is equal to the relative variance of the mean, multiplied by the number of observations ( $N$ ). For uncorrelated data, it



3.6

Figure 3.6 Effect of probit transformation ( The normal curve is transformed to a straight line when the ordinates are measured on a linear scale of probits instead of percentages after Finney, 1952 )

is equal to 1. Eq. 54 is not satisfied for averages of autocorrelated data. For averages of two values

$$s^2(\bar{x}_2) = \frac{s^2(x_t + x_{t+1})}{2} = (1/4) [s^2(x_t) + s(x_t, x_{t+1}) + s(x_{t+1}, x_t) + s^2(x_{t+1})] \quad (55)$$

where  $s(x_t, x_{t+1})$  and  $s(x_{t+1}, x_t)$  are covariances. If the series is stationary,

$$s^2(\bar{x}_2) = (1/2) s^2(x) [1 + r_1] \quad (56)$$

or

$$F_2 = (1 + r_1) \quad (57)$$

where  $r_1$  is the first autocorrelation coefficient (Agterberg, 1974b)

A similar expression can be developed for  $s^2(\bar{x})$ . If  $y = \bar{x}$  represents the mean of  $N$  values then,

$$s^2(y) = s^2(x) \left[ 1/N + (2/N^2) \sum_{j=1}^N (N-j) r_j \right] \quad (58)$$

These variances for the average values of successive samples were calculated by a FORTRAN program which was prepared by the author. (FORTRAN, Appendix I).

Agterberg (1974b) has shown that Eq. 58 becomes approximately

$$F_1 = 1 - c + (2c/a) \left\{ 1 + (e^{-al} - 1) / al \right\} \quad (59)$$

where the autocorrelation coefficient  $r(1)$  or  $r_1$  changes with distance or "lag"  $l$ . With  $\lim_{l \rightarrow \infty} r_1 = 1 - c + 2c/a$  for large values of  $l$ . The  $a$  and  $c$  estimated from the correlogram can be used for Eq. 59.

A theoretical curve based on the exponential model can be constructed using the variance values plotted against lag. This allows a comparison between experimental and theoretical values.

In an early application of the so called discrete Gaussian model, Matheron (1974) has considered the following distribution model for binary patterns.

$$X = \Phi \left[ (RZ - b) (1 - R^2)^{-1/2} \right] \quad (60)$$

In Eq. 60,  $X$  is a random variable generated from a binary variable  $X_0$ . The variable  $Z$  has a standard normal distribution with zero mean and unit variance,  $\Phi(u)$  represents the fractile of the cumulative normal distribution function with,

$$\Phi(u) = \frac{1}{(2\pi)^{1/2}} \int_{-\infty}^u \exp(-x^2/2) dx \quad (61)$$

If  $m$  represents the mean value of  $X$  for the study region, the constant  $b$  in Eq. 60 is defined as,

$$\Phi(b) = 1 - m \quad (62)$$

The constant  $R$  in Eq. 60 represents the correlation coefficient between the Gaussian random variable  $Z$  of Eq. 60 and a Gaussian variable  $Z_0$  which corresponds to the binary variable  $X_0$  which assume a value equal

to one for presence and a value equal to zero for absence of the feature which proportion values are measured. Agterberg (1981) demonstrated that:

$$X = \Phi \left[ \frac{RZ - b}{(1 - R^2)^{-1/2}} \right] = 1 - \Phi(b) - \phi(b) \sum_{j=1}^{\infty} R^j \frac{H_j(Z) H_{j-1}(b)}{(j!)} \quad (63)$$

This is equivalent to assuming that the probit of  $X$  satisfies a normal distribution. The variance of  $X$ ,

$$s^2 = \phi^2(b) \sum_{j=1}^{\infty} R^{2j} [H_{j-1}(b)]^2 \quad (64)$$

where,

$$\phi(u) = (2\pi)^{-1/2} \exp(-u^2/2) \quad (65)$$

and  $H_j(u)$  denotes a standardized Hermite polynomial.

### 3.4.1 Hermite polynomials

These satisfy,

$$H_j(u) = \left( u^j - \frac{j^{(2)}}{2 \cdot 1!} u^{j-2} + \frac{j^{(4)}}{2^2 \cdot 2!} u^{j-4} - \frac{j^{(6)}}{2^3 \cdot 3!} u^{j-6} + \dots \right) / (j!)^{1/2} \quad (66)$$

where

$$j^{(t)} = j(j-1)(j-2)\dots(j-t+1)$$

The first twelve standardized Hermite polynomials are:

$$H_0(u) = 1,$$

$$H_1(u) = u$$

$$H_2(u) = (u^2 - 1)(2)^{-1/2}$$

$$H_3(u) = (u^3 - 3u)(6)^{-1/2}$$

$$H_4(u) = (u^4 - 6u^2 + 3)(24)^{-1/2}$$

$$H_5(u) = (u^5 - 10u^3 + 15u)(120)^{-1/2},$$

$$H_6(u) = (u^6 - 15u^4 + 45u^2 - 15)(720)^{-1/2},$$

$$H_7(u) = (u^7 - 21u^5 + 105u^3 - 105u)(5040)^{-1/2},$$

$$H_8(u) = (u^8 - 28u^6 + 210u^4 - 420u^2 + 105)(40320)^{-1/2},$$

$$H_9(u) = (u^9 - 36u^7 + 378u^5 - 1260u^3 + 945u)(362880)^{-1/2},$$

$$H_{10}(u) = (u^{10} - 45u^8 + 630u^6 - 3150u^4 + 4725u^2 - 945)(3628800)^{-1/2},$$

$$H_{11}(u) = (u^{11} - 55u^9 + 990u^7 - 6930u^5 + 17325u^3 - 10395u)(39916800)^{-1/2},$$

$$H_{12}(u) = (u^{12} - 66u^{10} + 1485u^8 - 13860u^6 + 51975u^4 - 62370u^2 + 10395)(47900160)^{-1/2}.$$

The variances were calculated from these twelve Hermite polynomials with the FF FORTRAN computer program which is listed in Appendix I. For preparing the FF program use was made of an unpublished FORTRAN computer program by F.P. Agterberg called the HERMITE program in which Hermite polynomials until the 11<sup>th</sup> degree only were used.

### 3.4.2 Mean and variance relationships

The values of means and their corresponding fractile ( = probit - 5) values were taken from Table IX of Fisher and Yates ( 1963 ) for probits of a normal distribution and are shown in Table 3.1

TABLE 3.1 Input data for the FF FORTRAN program

m = mean	b = probit - 5	$\Phi(b)$ = probability
0.000001	4.75350	0.00000490
0.000010	4.26510	0.00004480
0.000100	3.71900	0.00039590
0.001000	3.09020	0.00336950
0.010000	2.32600	0.02667380
0.020000	2.05400	0.04839320
0.030000	1.88100	0.06801560
0.040000	1.75100	0.08612650
0.050000	1.64500	0.10311090
0.060000	1.55500	0.11908110
0.070000	1.47600	0.13422660
0.080000	1.40500	0.14868130
0.090000	1.34100	0.16233740
0.100000	1.28200	0.17540180

Probit values ( u ) can be used in Eq. 65 which is the prob - normal distribution for calculating the probability values as shown table 3.1. Table 3.1 was used as input for the FF FORTRAN program.

The value of R can be derived from the mean (  $\bar{x}$  ) and variance (  $s^2$  ) by using Eq. 64 or it can be estimated from mean and variance

graphs or tables as shown in Appendix II and III.

First, the means and variance relationships are plotted as shown in Appendix II. Next the vertical axis was changed to represent variance/mean (  $1.0 - \text{mean}$  ) in accordance with a suggestion made by Tukey ( 1984 ) in a discussion of a paper by Agterberg (1984b ). This quantity was plotted against the mean in Appendix II. Finally, the variance and mean relationships are examined in detail in Appendix II by plotting variances against logarithmic means. Appendix II. All graphs in Appendix II were plotted by the author using SAS graphic software.

The R values are tabulated with respect to their means and variances, their mean and [ variance/ ( mean (  $1.0 - \text{mean}$  ) )] and their logarithmic means and variances in Appendix III.

The frequency distribution of X in Eq. 60 becomes a straight line, if the function is plotted on normal probability paper. The ordinate represents the cumulative frequency percentage. A straight line is also obtained approximately by plotting the probit of the percentage value on normal probability paper and connecting the probits with one another.

### 3.4.3 Sampling errors and binomial distribution

If a variable X can assume a discrete set of values  $x_1, x_2, \dots, x_k$  with respective probabilities  $p_1, p_2, \dots, p_k$  when  $p_1 + p_2, \dots, p_k = 1$ , then this distribution is a discrete probability distribution. This is because X has been defined as the function  $P(x)$  which has the

respective values  $p_1, p_2, \dots, p_k$  for  $X = x_1, x_2, \dots, x_k$  is called the probability function or frequency function of  $X$ . It is binomial if it is represented by

$$P(x) = {}^n C_x p^x q^{n-x} = \frac{n!}{x! (n-x)!} p^x q^{n-x} \quad (67)$$

where  $q = 1 - p$

If the distribution of microfossil species was a binomial distribution model, the binomial mean is

$$\mu = np \quad (68)$$

where  $n$  the total of counts and  $p$  is the mean of proportion. The binomial standard deviation is,

$$\sigma = (n.p.q)^{1/2} \quad (69)$$

The binomial error can be defined as  $\sigma_B = \frac{100}{n} \sigma$ . The white noise ( see next section ) has variance given by,

$$\sigma_N^2 = (1 - c) s^2 \quad (70)$$

and standard deviation

$$\sigma_N = (\sigma_N^2)^{1/2} \quad (71)$$

The binomial error and white noise error can be compared with each other. They would be equal to one another if all white noise is due to sampling.

#### 3.4.4 Signal - plus - noise model

If a time series is weakly stationary and consists of two components; a systematic component ( signal ) and white noise component, a signal-plus- noise model can be applied.

The concept for models of this type is that a series of observed values  $x_k$  is the sum of the series signal (  $s_t$  ) and white noise (  $a_t$  ) with  $x_t = s_t + a_t$ . In geochemical applications, the signals are systematic variations in element concentration of a limited spatial extent. The signal may satisfy an autoregression process model ( AR ). The noise consists of uncorrelated values.

The corresponding three autocovariance functions are  $C_{xx}(l)$ ,  $C_{ss}(l)$  and  $C_{aa}(l)$  respectively, with

$$C_{aa}(l) = 0 \text{ if } l \neq 0 \text{ and } C_{xx}(l) = C_{ss}(l) + C_{aa}(l) \quad (72)$$

Because the noise is independent of the signal,

$$C_{sa}(l) = C_{as}(l) = 0$$

$$\text{Also } C_{xs}(l) = C_{sx}(l) = C_{ss}(l)$$

If the data X have variance or  $C_{xx}(0) = 1$ , then

$$\rho_{xx}(l) = C_{xx}(l)$$

However in general;  $\rho_{ss}(l)$  and  $\rho_{aa}(0) = 1 > C_{aa}(0)$ . The value c can be defined as  $C_{ss}(0) = c$  where c is a constant with  $0 \leq c \leq 1$ .

Suppose that

$$C_{ss}(l) = c e^{-al} \quad (73)$$

Agterberg ( 1978 ) has shown that the signal then satisfies a Markov process of the first order (  $X_t = \phi_1 X_{t-1} + a_t$  ) with

$$\rho_{xx}(1) = c e^{-a1} \quad \text{if } 1 = 0 \quad (74)$$

$$\rho_{xx}(0) = 1$$

If the noise is absent,  $c = 1$ , or if the series consists of noise only, then  $c = 0$ .

## CHAPTER 4 : Application of statistical and geostatistical models to middle - upper Jurassic foraminifers in Tojeira-sections, Monte junto area, Portugal.

### 4.1- Introduction

The following application studies are designed to show how to construct a geostatistical model. Tables 1.1, 1.2, 1.3 and 1.4, which show the data sets of Tojeira 1 and Tojeira 2 section, are used to calculate the autocorrelation coefficients with the SAS package called ARIMA.

Ideally, the number of observations should be a minimum of about 30 in order to proceed with the ARIMA procedure. This usually allows sufficient degrees of freedom for adequate identification and estimation. Occasionally, some analyses may use fewer than 30 observations, but the results then must be interpreted cautiously (see SAS Time Series User Guide Version 5).

Box and Jenkins ( 1976 ) suggest that about 50 observations are required for an ARIMA procedure. However, Jenkins ( 1979, p.64 ) presents an example of a ( multivariate ) ARIMA procedure based on only 14 observations. The key is not necessarily the total number of observations but rather the amount of " statistical noise " in the data. If the noise factor ( the variance of the random shocks or white noise ) is small, it may be possible to extract enough information from relatively few observations in order to develop a useful ARIMA procedure.

Table 1.1 is the data set of Tojeira 1 that was used with the ARIMA procedure to calculate the autocorrelation coefficients and residuals. These autocorrelation coefficients were plotted against lags. Printout 1 ( see appendix I ) shows that 3 species of the 31 samples have autocorrelation coefficients which are significantly different from zero. These three species ( Eoguttulina sp., Epistomina mosquensis and Ophthalmidium strumosum ) were examined more closely.

The SAS variable name should be less than 8 characters so A2 was used in place of Eoguttulina sp. The A indicates it is from Tojeira 1 and the 2 indicates it is from the second row of the Table 1.1 The B indicates it is from Tojeira 2 section ( Table 1.2 )

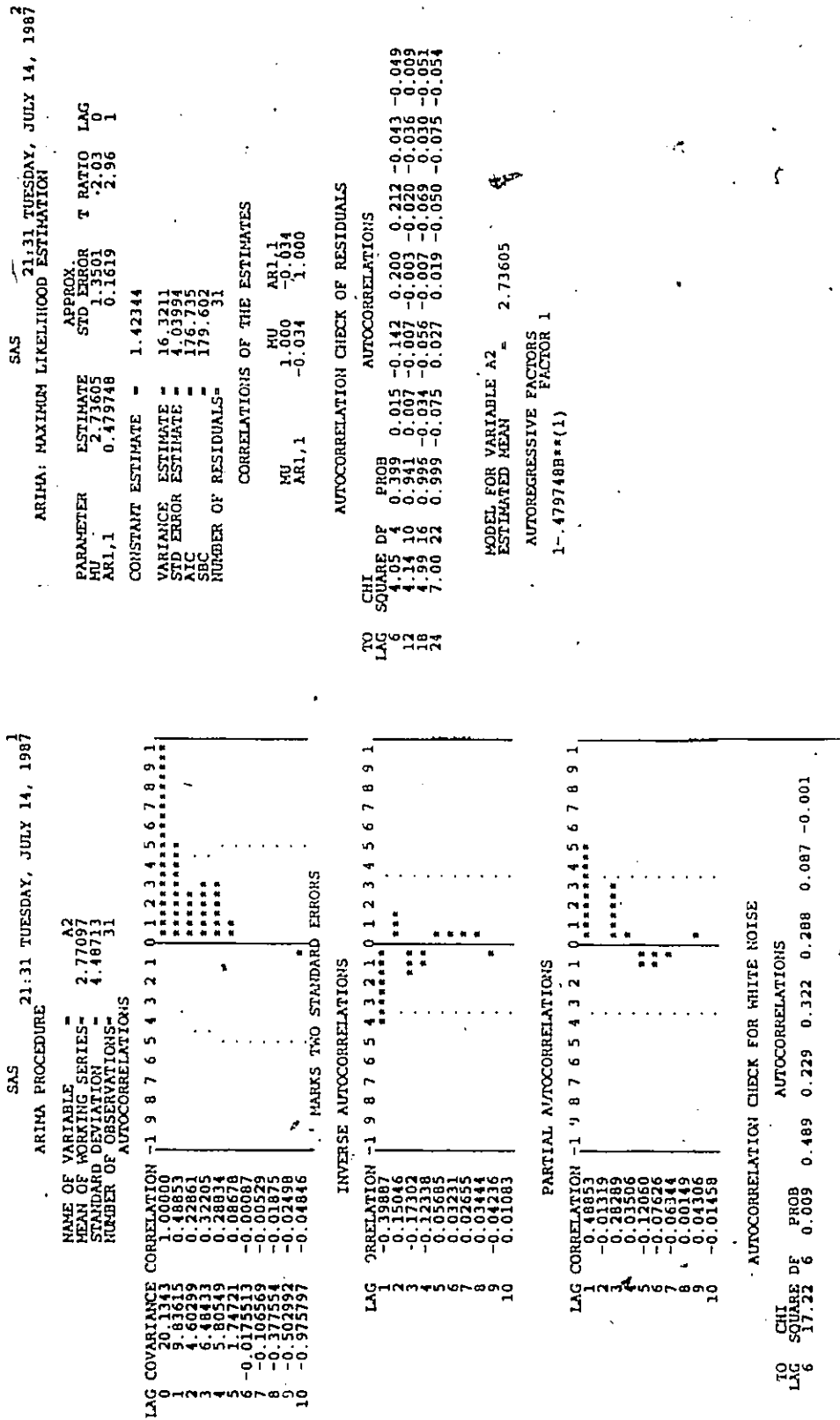
#### 4.2 - Tojeira 1 section

##### 4.2.1 - EOGUTTULINA SP.

The number of observations is 31 (  $N = 31$  ) as shown in Table 1.1 so that there are approximately  $31/4 = 8$  autocorrelations and also 8 approximately partial autocorrelations. These are useful primarily for identifying the AR order of the ARIMA procedure.

##### 4.2.1.1 Time series analysis - searching for a good model to fit the data set

Figure 4.1 Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to A2(Eoguttulina sp.)



The estimated a.c.f and p.a.c.f are shown in Fig. 4.1. The estimated autocorrelation function decays toward zero at the positive side, while the theoretical p.a.c.f has a single positive asterisk at lag 1 when  $q = 1, p = 0, d = 0$  for ARIMA ( q d p ) or ARIMA ( 1 0 0 ) or AR ( 1 ). The AR ( 1 ) model can be selected, tentatively, for processing the data,

$$X_t = C + \hat{\phi}_1 X_{t-1} + a_t \quad (A)$$

Model A has two parameters requiring estimation. Fig. 4.1 results from fitting model ( A ) where  $C = 1.423$  and  $\hat{\phi}_1 = 0.479$ . The absolute value of  $\hat{\phi}_1$  is less than 1. This model is stationary. The absolute t - value ( from Fig. 4.1 ) is greater than 2.0 and this estimated coefficient value is different from zero at the 5% significance level.

The random shocks (  $a_t$  ) are independent. If the shocks in a given model are correlated, the model must be reformulated, because it does not fully capture the statistical relationship among the X's. Fig.4.1 shows the residual a.c.f for model ( A ). The random shock t - value is less than 1.25 for lag 1. The random shock at this lag is independent. The absolute t - values and the chi - squared statistic are all relatively small. Consequently, it may be concluded that all random shocks are uncorrelated and model ( A ) is statistically adequate as a model with  $X_t = 1.423 + 0.479X_{t-1} + a_t$ . This result is shown on Fig. 4.1 with B notation.

Figure 4.2 Printout of estimated a.c.f.'s and P.a.c.f.'s after fitting the MA(1) time series model to A2( Eoguttulina sp.)

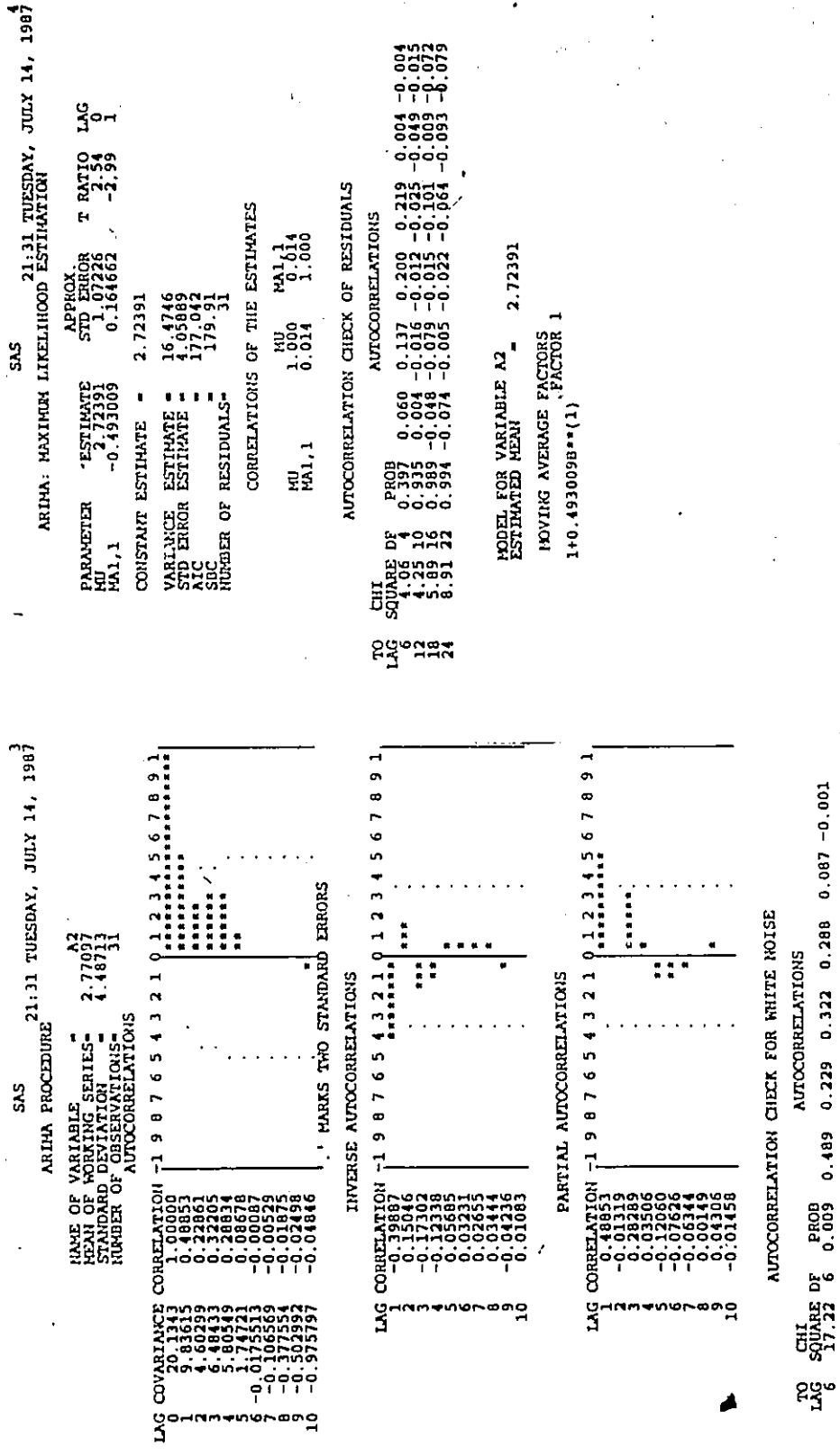


Fig. 4.2 shows the estimated a.c.f and p.a.c.f for the MA ( 1 ) process where  $\theta_1 < 0$  ( $\theta_1$  is negative). The asterisks in the a.c.f indicate positive values. The a.c.f decays exponentially, starting on the positive side, and the p.a.c.f also damps out exponentially starting on the positive side. The estimated a.c.f drops to nearly zero after lag 1. This model is written as,

$$X_t = C - \theta_1 a_{t-1} + a_t \quad (B)$$

where  $q = 0$ ,  $p = 1$  and  $d = 0$  for ARIMA ( p d q ) or MA( 1 ). The B model is a pure MA process or white noise series, and no further checks are required ( Fig. 4.2 ).

Consider the estimated a.c.f in Fig. 4.3. Only the first autocorrelation is significantly different from zero at the 5% significance level. Only the first asterisk in the a.c.f falls outside the 95% confidence belt on the printout. The position of this belt is based on Bartlett's approximation for the standard error ( Eq. 19 )

$$s(r_1) = (1 + 2.0)^{1/2} (31)^{-1/2} = 0.179$$

used in the 95% confidence interval  $= \rho_1 \pm 2 s ( r_1 )$ . For the null hypothesis  $H_0 : \rho_1 = 0$ , this interval is  $\pm 2 s ( r_1 ) = \pm 0.358$ .

Because  $r_1 > 2s ( r_1 )$  the asterisk in the a.c.f falls outside the belt on the printout.

Figure 4.3 Printout of the ARMA(1,1) time series model for A2( Eoguttulina sp.)

SAS 9:28 THURSDAY, APRIL 23, 1987

ARIMA PROCEDURE NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES = 2.77097  
 STANDARD DEVIATION = 4.48713  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

SAS 9:28 THURSDAY, APRIL 23, 1987  
 ARIMA: MAXIMUM LIKELIHOOD ESTIMATION

PARAMETER	ESTIMATE	APPROX. STD ERROR	T RATIO	LAG
MA, 1	-0.278989	0.366717	-0.76	1
AR, 1	0.754803	0.365735	0.70	1

CONSTANT ESTIMATE = 2.0183  
 VARIANCE ESTIMATE = 16.8526  
 STD ERROR ESTIMATE = 4.10519  
 AIC = 178.664  
 SDC = 182.966  
 NUMBER OF RESIDUALS = 31

CORRELATIONS OF THE ESTIMATES  
 MA, 1 AR, 1  
 MA, 1 1.000 -0.022 -0.035  
 AR, 1 -0.022 1.000 0.867  
 AR, 1 -0.035 0.867 1.000

AUTOCORRELATION CHECK OF RESIDUALS

MODEL FOR VARIABLE A2  
 ESTIMATED MEAN = 2.70841

AUTOREGRESSIVE FACTORS  
 1 - .2548030\*\*(1)

MOVING AVERAGE FACTORS  
 \*\*0.2789898\*\*(1)

CHI SQUARE DF PROB  
 1 22 3 0.378 -0.003 -0.009 0.213 0.200 -0.025 -0.021  
 2 3 36 9 0.948 0.010 -0.014 -0.007 -0.020 -0.041 0.001  
 3 4 56 15 0.995 -0.041 -0.066 -0.005 -0.088 0.021 -0.059  
 4 6 75 21 0.999 -0.075 0.014 -0.002 -0.056 -0.078 -0.062

LAG COVARIANCE CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1

LAG CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1

PARTIAL AUTOCORRELATIONS  
 LAG CORRELATION -1 9 8 7 6 5 4 3 2 1 0 1 2 3 4 5 6 7 8 9 1

AUTOCORRELATION CHECK FOR WHITE NOISE  
 LAG SQUARE DF PROB  
 6 17.22 6 0.009 0.489 0.229 0.322 0.268 0.087 -0.001

LAG	COVARIANCE CORRELATION	CORRELATION	PARTIAL AUTOCORRELATION
1	1.00000	0.48853	0.48853
2	0.48853	0.22861	0.22861
3	0.22861	0.32205	0.32205
4	0.32205	0.28834	0.28834
5	0.28834	0.08678	0.08678
6	0.08678	-0.00087	-0.00087
7	-0.00087	-0.00529	-0.00529
8	-0.00529	-0.01675	-0.01675
9	-0.01675	-0.02498	-0.02498
10	-0.02498	-0.04846	-0.04846

LAG	CORRELATION
1	0.48853
2	0.22861
3	0.32205
4	0.28834
5	0.08678
6	-0.00087
7	-0.00529
8	-0.01675
9	-0.02498
10	-0.04846

LAG	PARTIAL AUTOCORRELATION
1	0.48853
2	-0.01319
3	0.28289
4	0.03506
5	-0.12060
6	-0.07626
7	-0.06344
8	0.00149
9	0.04306
10	-0.01458

AUTOCORRELATION CHECK FOR WHITE NOISE

LAG SQUARE DF PROB

6 17.22 6 0.009 0.489 0.229 0.322 0.268 0.087 -0.001

For  $r_2$ ,  $s(r_2) = (1 + 2r_1^2)^{1/2} 31^{-1/2} = 0.218$

The 95% interval for  $r_2$  is  $\rho_2 \pm 2s(r_2)$ . Setting  $\rho_2 = 0$  gives  $\pm 2s(r_2) = 0.436$  and  $r_2 < 0.436$

The autocorrelations decrease gradually to zero so that the mean of the series is probably stationary. Since  $\hat{\phi}_1 = 0.259$  and  $|\hat{\phi}_1| < 1$  then according to Table 2.1 this series is stationary. An AR procedure seems appropriate because the a.c.f decays toward zero rather than cutting off sharply to zero. A decaying a.c.f is also consistent with a mixed (ARMA) procedure. A model of this working series is an ARMA(1, 1) (see Fig. 4.3). ARMA(1, 1) can be written as

$$X_t = C + \phi_1 X_{t-1} - \theta_1 a_{t-1} + a_t \quad (AB)$$

For model (AB), estimated coefficients from Fig. 4.3 are;

$$\hat{\phi}_1 = 0.255$$

$$\hat{\theta}_1 = -0.279$$

$$\hat{C} = \hat{\mu} (1 - \hat{\phi}_1) = 2.018$$

If the p.a.c.f was tested, the standard error according to Eq.22 would be

$$s(\hat{\phi}_{11}) = 31^{-1/2} = 0.197$$

The t - statistic is calculated using Eq. 23 used to test the null hypothesis  $H_0 : \phi_{11} = 0$

$$t_{\hat{\phi}_{11}} = \frac{0.49 - 0}{0.197} = 2.487$$

The absolute value of the t - statistic is greater than 2.0. The  $\hat{\phi}_{11}$  is different from zero at the 5% significance level so the null hypothesis  $\phi_{11}$  is rejected.

Let  $k = 2$  and test the null hypothesis  $H_0 : \phi_{22} = 0$

$$t_{\hat{\phi}_{22}} \text{ - statistic} = \frac{-0.01319 - 0}{0.179} = -0.07$$

The absolute value of the t value is less than 2.0 so the null hypothesis is acceptable and  $\phi_{22} = 0$ .

Fig. 4.3 shows the result of estimating the ARIMA procedure of Eoguttulina sp. The  $\phi$  and  $\mu$  were estimated using the likelihood criterion. In Fig.4.3  $\hat{\phi}_1 = 0.255$  and  $\hat{\mu} = 2.708$  and the estimated constant is found to be  $C = 2.018$ . This model satisfies the stationarity requirement  $|\hat{\phi}_1| < 1.0$ . The t - statistic for  $H_0 : \phi_1 = 0$  is as follows,

$$t = \frac{0.2548 - 0}{0.365} = 0.698 = 0.70 \text{ ( as in Fig.4.3 )}$$

The t value ( 0.70 ) is less than 2.0, so the null hypothesis is acceptable for  $\phi_1 = 0$ . According to the result of this statistical significance testing, the coefficient quality is not good or, as shown in Fig. 4.3,  $\hat{\phi}_1$  and  $\hat{\theta}_1$  are not statistically significant.

If Fig. 4. 3 was statistically adequate, the random shocks  $a_t$  could

be tested for independence using the residuals  $\hat{a}_t$  from the estimated equation. The residuals are estimates of the random shocks and these shocks are assumed to be statistically independent. The estimated a.c.f of the residuals are used to test whether the shocks are independent. The 8 of the 31 residual autocorrelations are examined using the  $t_b$  - test. Using Eq. 33, we obtain

$$s [r_1 (\hat{a}_1)] = (1 + 2.0)^{1/2} 31^{-1/2} = 0.179$$

$$t_{b_1} = \frac{-0.003 - 0}{0.179} = -0.015$$

$$s [r_2 (\hat{a}_2)] = (1 + 2 (0.01)^2)^{1/2} 31^{-1/2} = 0.1796$$

$$t_{b_2} = \frac{0.01 - 0}{0.1796} = 0.0556$$

The absolute value of the residual a.c.f  $t_b$  - value is less than 1.25 at lags 1, and 2. Furthermore, according to the  $X^2$  (chi - squared) test (Eq. 34), if 6 residual autocorrelations are assumed, then  $K = 6$  and three parameters will be estimated ( $\phi$ ,  $\theta$ , and  $\mu$ ) so that  $m = 3$ . Therefore, there are  $K - m = 3$  degrees of freedom. From Fig. 4.3 this value is 3.22, and according to the chi - squared table the critical value with degrees of freedom equal to 3 at the 5% significance level is 7.81. For 6 degrees of freedom, the value determined from Fig. 4.3

is 3.36 and the corresponding table value is 16.9. The calculated values are less than the critical values. The shocks appear to be independent according to the  $t_b$  - test as well as the chi - squared test.

The model ( AB ) may be a good model for processing the data set. However, the estimated coefficients are not statistically significant.

Model ARMA ( 1, 2 ) (see the Fig. 4.4 ) and Model ARMA ( 2, 1 ) ( see Fig. 4.5 ) have estimated coefficients with absolute  $t$  - values less than 2.0. They are not significantly different from zero at the 5% significance level. The selected time series model is AR ( 1 ) or model ( A ). The working data set has a good fit with the AR ( 1 ) model. If the AR ( 1 ) model and ARMA ( 1, 1 ) model are essentially the same in all other respects, the AR ( 1 ) model would be selected because it has one less coefficient to estimate.

#### 4.2.1.2 SPECTRA procedure

In Fig. 4.6, illustrating the SPECTRA process, it is shown that the maximum periodogram ordinate is 168.265 and the periodogram ordinate total is 624.164, for  $M-1=15$  (31 samples have 16 periodogram ordinates ).

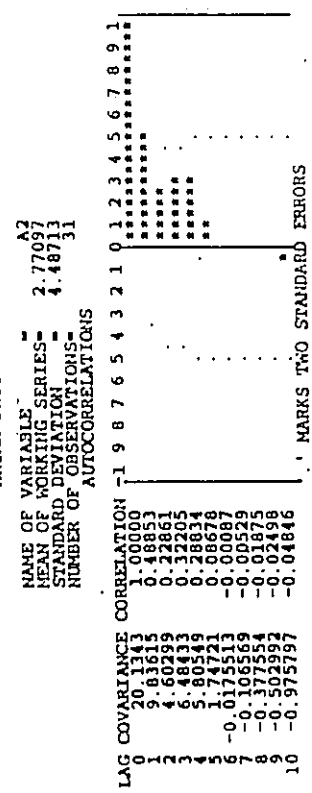
$$\text{Kappa test statistic : } \frac{15 \cdot 168.265}{624.164} = 4.0438$$

The table value is 5.019 for the 5% significance level . The calculated value is less than the table value. Consequently, the null hypothesis is

Figure 4.4 Printout of the ARMA(1,2) time series model for A2( Eoguttulina sp.)

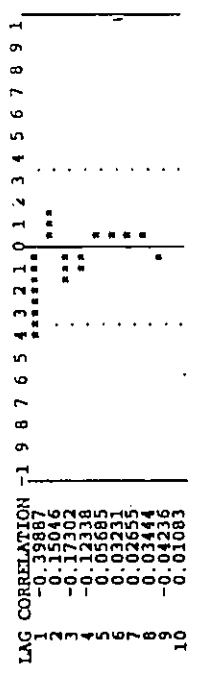
SAS 21:31 TUESDAY, JULY 14, 1987  
 ARIMA PROCEDURE  
 NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES = 2.77097  
 STANDARD DEVIATION = 4.48713  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS  
 DID NOT CONVERGE AFTER 15 ITERATIONS.  
 ARIMA: MAXIMUM LIKELIHOOD ESTIMATION  
 SAS 21:31 TUESDAY, JULY 14, 1987

LAG COVARIANCE CORRELATION

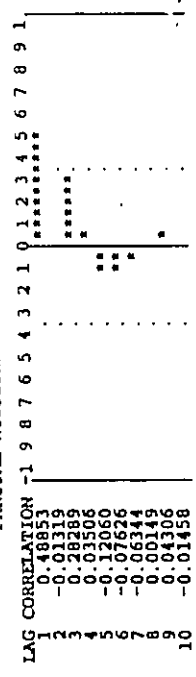


MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS



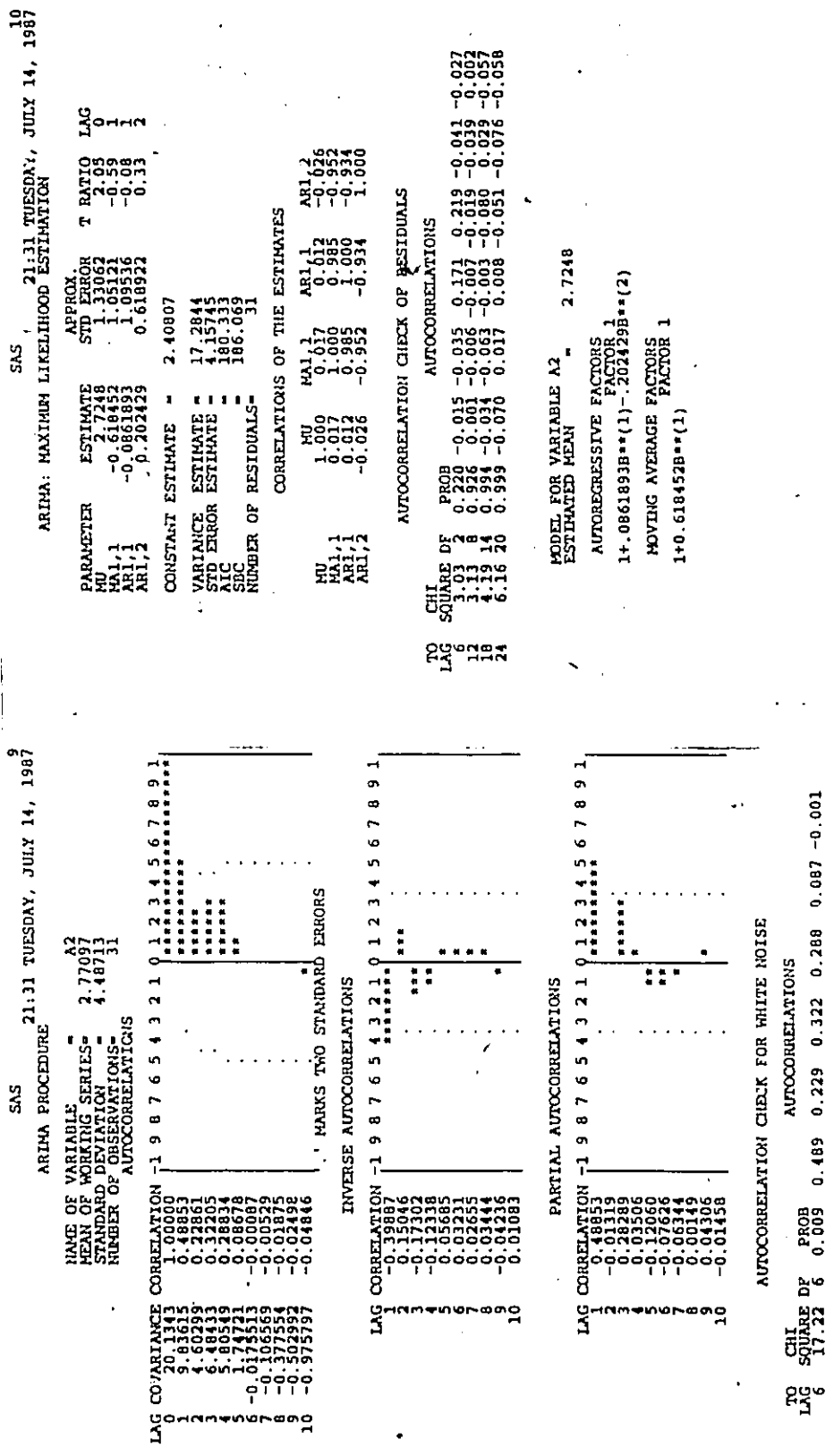
PARTIAL AUTOCORRELATIONS



AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	PROB	AUTOCORRELATIONS
LAG	SQUARE	DF	
6	17.21	6	0.009 0.489 0.229 0.322 0.288 0.087 -0.001

Figure 4.5 Printout of the ARMA(2,1) time series model for A2( Eoguttulina sp.)





rejected (the hypothesis of independence). The series is not pure white noise.

The SAS package program calculated the Kolmogorov - Smirnov test for 1% probability. The 5% test was calculated by the author using the equation:

$$\text{Test} : \frac{1.36}{(15)^{1/2}} = 0.351$$

Here 1.36 is a constant and 15 represents the number of the periodogram ordinates (Fig. 4.6). Furthermore, investigation of the white noise hypothesis was done by plotting the cumulative periodogram values against the lag values as shown in Fig. 4.7. The  $k / (15)^{1/2} \pm 0.351$  values were plotted against lag values in the same figure.

Fig. 4.7 shows that the series of Eoguttulina sp. is not white noise.

Statistical properties of the Eoguttulina sp. series are well known now so that the geostatistical model can be applied to it. All the required statistics could be calculated using the SAS package program.

The pattern of autocorrelation coefficients plotted against lag or correlogram shows that the trend is approximately exponential. If the vertical axis was given a logarithmic scale, the exponential curve becomes a straight line. The best fitting line with the constants a and c can be constructed. When estimated from the correlogram in Fig. 4.8, these values are,  $a = 0.24$  (a is the slope of best fitting line) and  $c = 0.76$ .

Eq. 47 becomes

# PERIODOGRAM

A2:EOGUTTULINA SP.

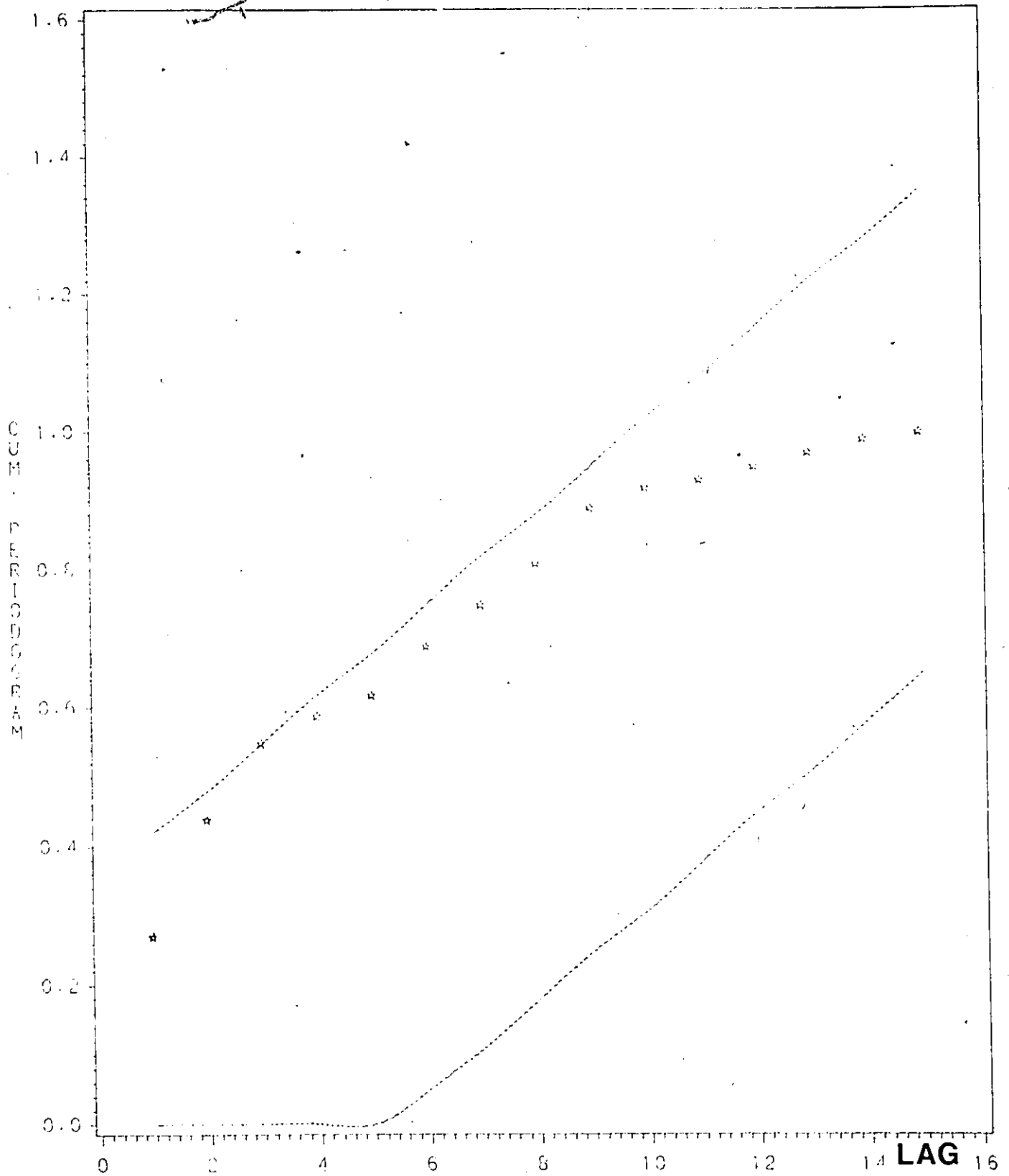
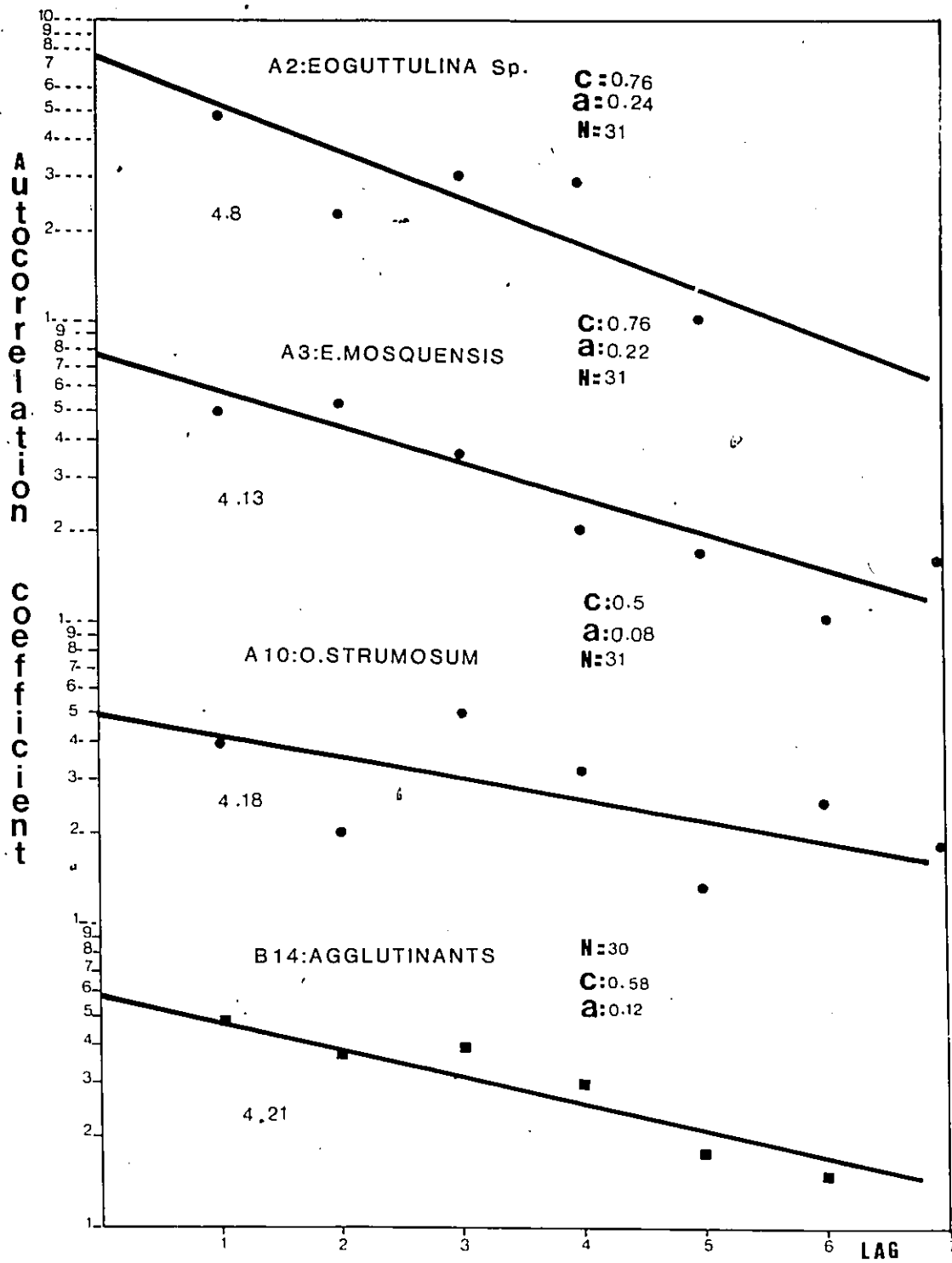


Figure 4.7 The periodogram for A2( Eoguttulina sp.).

Figure 4.8 The correlogram for A2( Eoquuttulina sp.) and the best-fitting exponential curve with  $a = 0.24$  and  $c = 0.76$

Figure 4.13 The correlogram for A3( E. mosquensis) with best - fitting exponential curve. From this correlogram,  $a = 0.22$  and  $c = 0.76$  were estimated

Figure 4.18 The correlogram for A10( O. strumosum) and the best-fitting exponential curve with  $a = 0.08$  and  $c = 0.50$



$$r(k) = 0.76e^{-0.24k}$$

This result can be used in Eq. 54.  $S^2(x_N)$  values which are variances for  $N$  adjacent samples, are calculated by the F FORTRAN program ( see Appendix I ). For a series of 31 values, there are 30 pairs of adjacent values, 29 sets of three adjacent values, etc.. Observed relative variances ( $F_N$ ) were calculated using values of  $S^2(x_N)$  with the results shown in Table 4.1.

TABLE 4.1 Theoretical and observed values of the relative variance of Eoguttulina sp.

k	$S^2(x_N)$	$F_N$	$F_L$
0	0	0	0.24
1	20.85	1	0.94
2	15.95	1.53	1.53
3	12.54	1.80	2.06
4	9.53	1.83	2.50
5	8.00	1.92	2.88
6	7.10	2.04	3.21
7	5.63	1.89	3.50

The value  $F_N$  represents the relative variance calculated according to Eq. 54. We have for example,

$$F_1 = \frac{1 \cdot 20.85}{20.85} = 1$$

$$F_2 = \frac{2 \cdot 15.95}{20.85} = 1.53$$

If the data were uncorrelated (white noise model), all  $F_N$  values would approximately be equal to 1. The  $F_N$  factors for  $N > 1$  are considerably larger than unity as predicted by Eq. 54 which is not valid for autocorrelated data. The standard deviation of the mean of 31 uncorrelated data would be

$$s(\bar{x}_{31}) = \frac{4.48}{\sqrt{31}} = 0.804$$

The population mean ( $\mu$ ) would be in the interval  $\bar{x} \pm 0.804 \cdot t_{0.975}$  with a probability of 95% or because  $t_{0.975}(30) = 2.06$

$$P(1.114 < \mu < 4.426) = 95\%$$

Because the data are autocorrelated, Eq. 54 must be used instead with

$$s(\bar{x}_{31}) = \sqrt{31} \cdot s(x) / \sqrt{F_{31}}$$

It means that the above estimate of  $s(\bar{x}_{31}) = 0.804$  must be corrected by the factor  $\sqrt{F_{31}}$ . But  $\sqrt{F_{31}}$  can not be estimated from the data by the direct method used, for example, to estimate  $F_2$ . However, by assuming that the series is stationary and consists of two components characterized by  $a$  and  $c$ , by approximation, we can use Eq. 59 to

calculate  $F_{31}$ . This gives

$$F_{31} = 1 - 0.76 + (2 * 0.76 / 0.24) [1 + (e^{-0.24 * 31} - 1) / 0.24 * 31]$$

or  $F_{31} = 5.72$ , and

$$s(x_{31}) = \sqrt{5.72 * 0.804} = 1.923$$

The above confidence interval for the mean must be replaced by

$$P(0.00 < \mu < 6.76) = 95\%$$

In fact, the lower limit of the confidence interval is negative, however the mean can never be a negative. According to Eq. 59,

$$F_0 = 1 - 0.76 = 0.24$$

$$F_1 = 0.24 + 6.33 [1 + (e^{-0.24 * 1} - 1) / 0.24 * 1] = 0.94$$

$$F_2 = 0.24 + 6.33 [1 + (e^{-0.24 * 2} - 1) / 0.24 * 2] = 1.53$$

(as above in Table 4.1).

The  $F_N$  and  $F_L$  variance values are plotted against lag as shown in Fig. 4.9. The correspondence between the experimental and theoretical values is good for this example.

The Eoguttulina data, pairs of Eoguttulina data and the sets of 4 adjacent Eoguttulina data were transformed with the probit transformation (Table 4.2) using the table value. Frequency distribution tables for these data were constructed as in Table 4.3.

Calculated cumulative frequency distributions were plotted on normal probability paper using upper class limits. They fall approximately on the straight lines (Chapter 2). Eq. 60 provides a theoretical straight line on the probability paper for percentage values of this type. The distribution for the 31 values of A2 (Eoguttulina sp.) has mean  $(\bar{x}) = 0.0277$  and variance  $s^2(x) = 0.002013$ . By using

# THEORETICAL CURVE

A2:EOGUTTULINA SP.

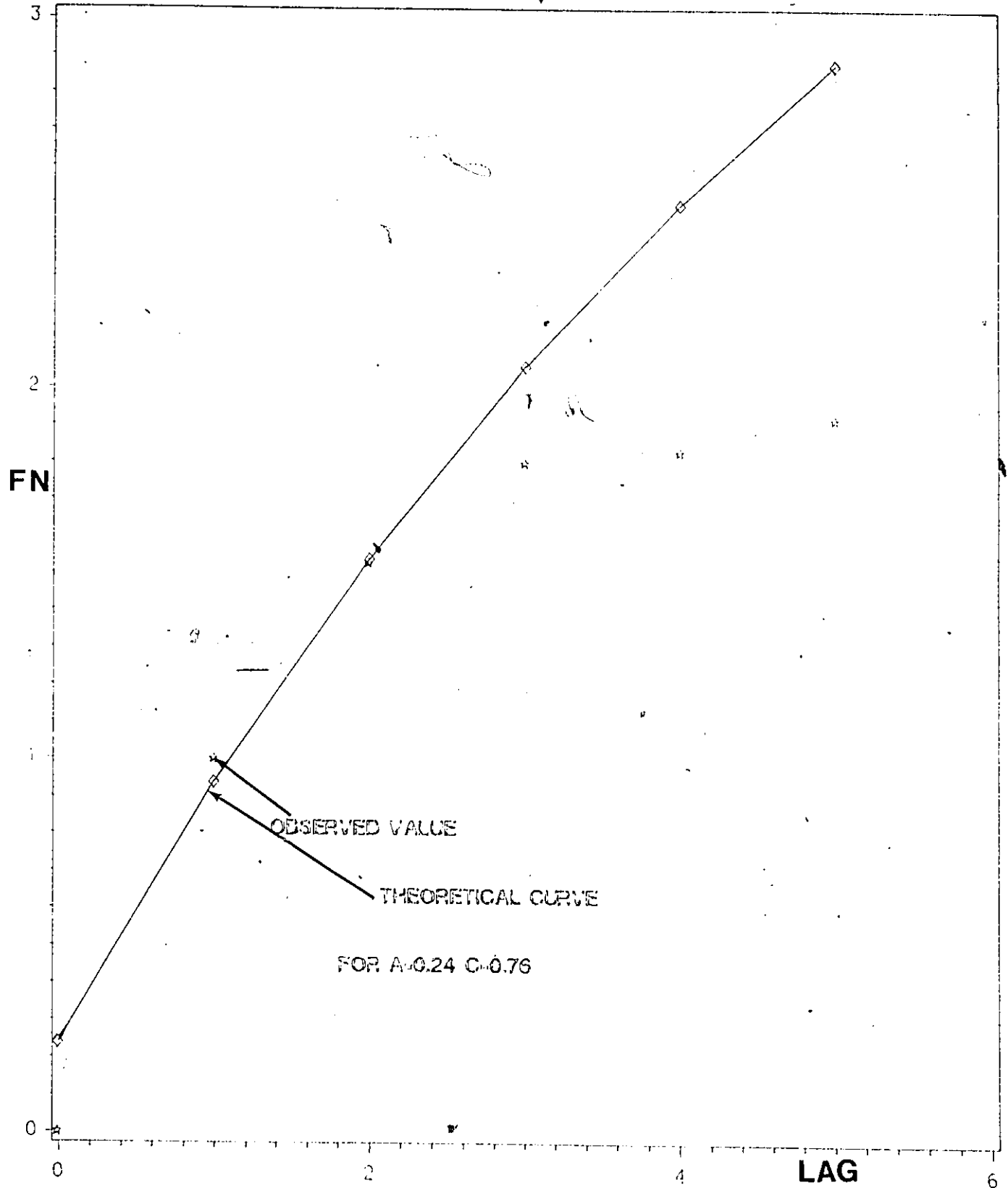


Figure 4.9 Theoretical curve for  $a = 0.24$  and  $c = 0.76$  in comparison with observed values of Eoguttulina sp.

Table 4.2 Probit transformation of A2 (Eoguttulina sp.) from the Tojeira 1 section.

Sample single %	probit	sample pairs %	probit	sample set of 4 %	probit
4.1	3.261				
14.8	3.955	9.45	2.686		
19.0	4.122				
5.4	3.393	12.0	3.835	10.825	3.764
1.8	2.903				
13.5	3.897	7.65	3.570		
2.2	2.986				
3.4	3.175	2.80	3.080	5.225	3.376
0.3	2.252				
2.6	3.057	1.45	2.816		
1.1	2.710				
1.1	2.710	1.1	2.710	1.275	2.766
1.3	2.774				
0.6	2.488	0.95	2.654		
1.5	2.830				
0.3	2.252	0.90	2.634	0.925	2.644
0.5	2.424				
1.1	2.710	0.80	2.591		

( continued )

Table 4.2 ( concluded )

0.3	2.252				
0.8	2.591	0.6	2.488	0.70	2.543
0.8	2.591				
0.8	2.591	0.8	2.591		
3.5	3.188				
1.5	2.830	2.5	2.424	1.65	2.510
0.6	2.488				
0.6	2.488	0.6	2.488		
0.9	2.634				
1.0	2.674	0.95	2.654	0.775	2.578
0.3	2.252				
0.0	2.187	0.15	1.916		
0.2	2.182				

Table 4.3 Cumulative frequency distribution of A2( Eoguttulina sp.) for 31, 15 and 7 samples.

N = 31 ( single observed values )

class interval	mid point	Freq.	Cum. freq	Freq %	Cumfreq %	plotting value
2.2 - 2.6	2.4	13	13	41.96	41.96	40.4
2.6 - 3.0	2.8	10	23	32.26	74.19	72.3
3.0 - 3.4	3.2	5	28	16.13	90.32	88.3
3.4 - 3.8	3.6	0	28	0.00	90.32	88.3
3.8 - 4.2	4.0	3	31	9.68	100.00	97.9

N = 15 ( pairs of observed values )

1.7 - 2.3	2.0	1	1	6.67	6.67	4.3
2.3 - 2.8	2.55	9	10	60.0	66.67	63.0
2.8 - 3.3	3.05	2	12	13.33	80.00	76.1
3.3 - 3.8	3.55	3	15	20.00	100.00	97.9

N = 7 ( 4 of set of observed values )

2.6 - 3.0	2.8	5	5	71.43	71.43	63.6
3.0 - 3.4	3.4	1	6	14.29	85.71	77.73
3.4 - 3.8	3.6	1	1	14.29	100.00	90.90

NOTE :

$$\text{Plotting value} = \frac{\text{cum. freq.} * 3 - 1}{(N * 3 + 1) / 100}$$

Appendix II or III,  $R = 0.34$  was found. From Eq. 62, it follows that

$$\Phi(b) = 1 - 0.0277 = 0.97229$$

The probit of 0.97229 is 6.916 ( see Table III ) and

$$b = 6.916 - 5 = 1.916$$

Eq. 60 becomes  $Y_1 = 0.3615z - 2.037$  using the above value for  $N = 31$ . This function is plotted on normal probability paper which is the same paper used for the probit transformed data ( Fig. 4.10 a ).

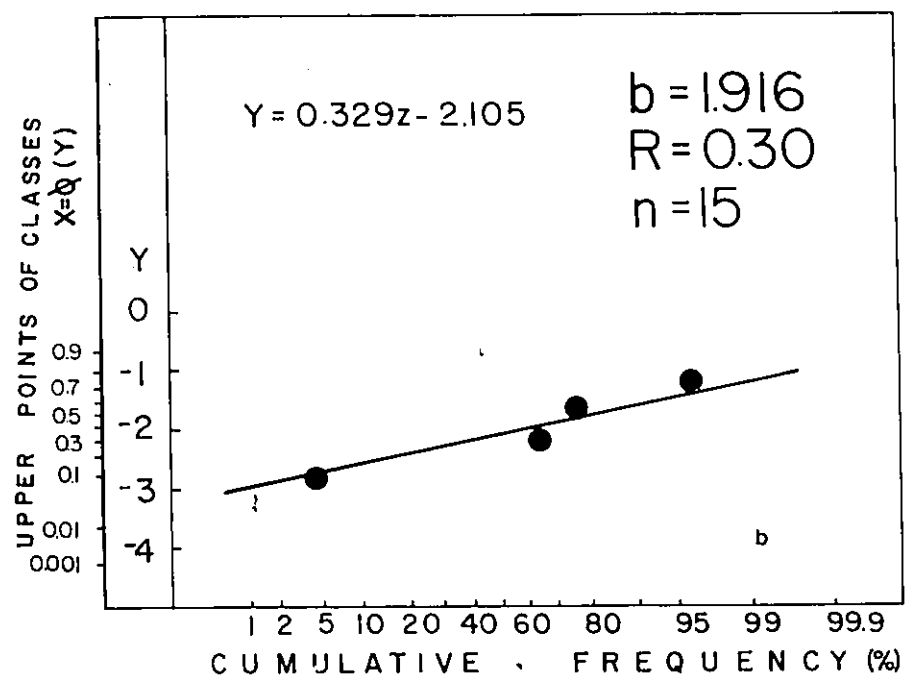
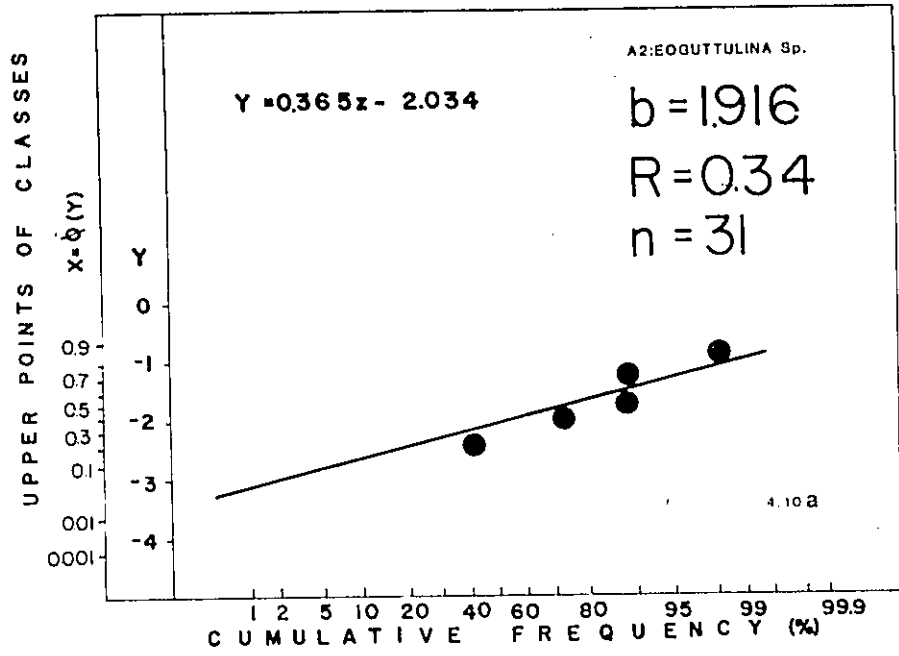
For the distribution of 15 successive values ( pairs of adjacent values ) of  $A_2$ ( Eoguttulina sp.), the mean is the same as that of 31 single values of  $A_2$ ( Eoguttulina sp.) and its variance. Using Eq. 54,

$$s^2(x_2) = \frac{F_2 s^2(x)}{2} = 0.001540$$

From Appendix II or III, it follows that  $R = 0.30$  with  $b = 1.916$  (  $b$  is dependent on the mean ). Eq. 60 becomes  $Y_2 = 0.329z - 2.105$ . This  $Y$  function is plotted on normal probability paper which is the same paper used before for the probit transformed data of the paired values of  $A_2$ ( Eoguttulina sp. ) Fig. 4.10b.

The distribution of the 7 successive relative values of Eoguttulina sp. ( sets of 4 adjacent measured values ) also has the same mean and  $b$  coefficient. Its variance is,

Figure 4.10a,b,c Experimental frequency distributions for 31, 15, and 7 values of  $A_2$  (Eoquuttulina sp.) plotted on two normal probability scales. The values of  $b$  and  $R$  were obtained from the means and variances. They determine the positions and slopes of the lines for probnormal distributions satisfying Eq.60



$$s^2(x_4) = \frac{(F_4 s^2(x))}{4} = 0.000916$$

By using Appendix II or III, it follows that  $R = 0.25$ . Eq. 60 becomes  $Y = 0.26z - 1.98$ . For  $Y_4$ , this function is shown graphically in the same manner as the other  $Y$  functions. The vertical scales are  $X = \phi(Y_n)$  in Fig. 4.10c. There is good agreement between the observed values and the theoretical distributions as can be seen in Fig. 4.10 a, b, c., which means that Eoguttulina sp. is a unique species. The distribution of this taxon could be of biostratigraphic value. It could be used for correlation between sections using the abundance data for this species as well as for correlation with other species within the same samples.

#### 4.2.2 EPISTOMINA MOSQUENSIS

As in section 4.2.1 the ARIMA procedure is used to calculate autocorrelation coefficients. The estimated a.c.f's for the AR ( 1 ) model is seen in Fig. 4.11 and the model shown in Fig. 4.12. ARMA ( 1, 2 ) may be good time series models for the data set of Epistomina mosquensis. The AR ( 1 ) was tested and the estimated a.c.f for this model is shown in Fig. 4.11. The first two autocorrelations are significantly different from zero, at the 5% level of significance. The first and second asterisks in the a.c.f extend beyond the 95% confidence belt of two standard errors. Since  $\hat{\phi}_1 > 0$  and the absolute value of  $\phi_1 < 1$  then, AR ( 1 ) is stationary. The AR process has theoretical a.c.f's

Figure 4.11 Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to A3(E. mosquensis)

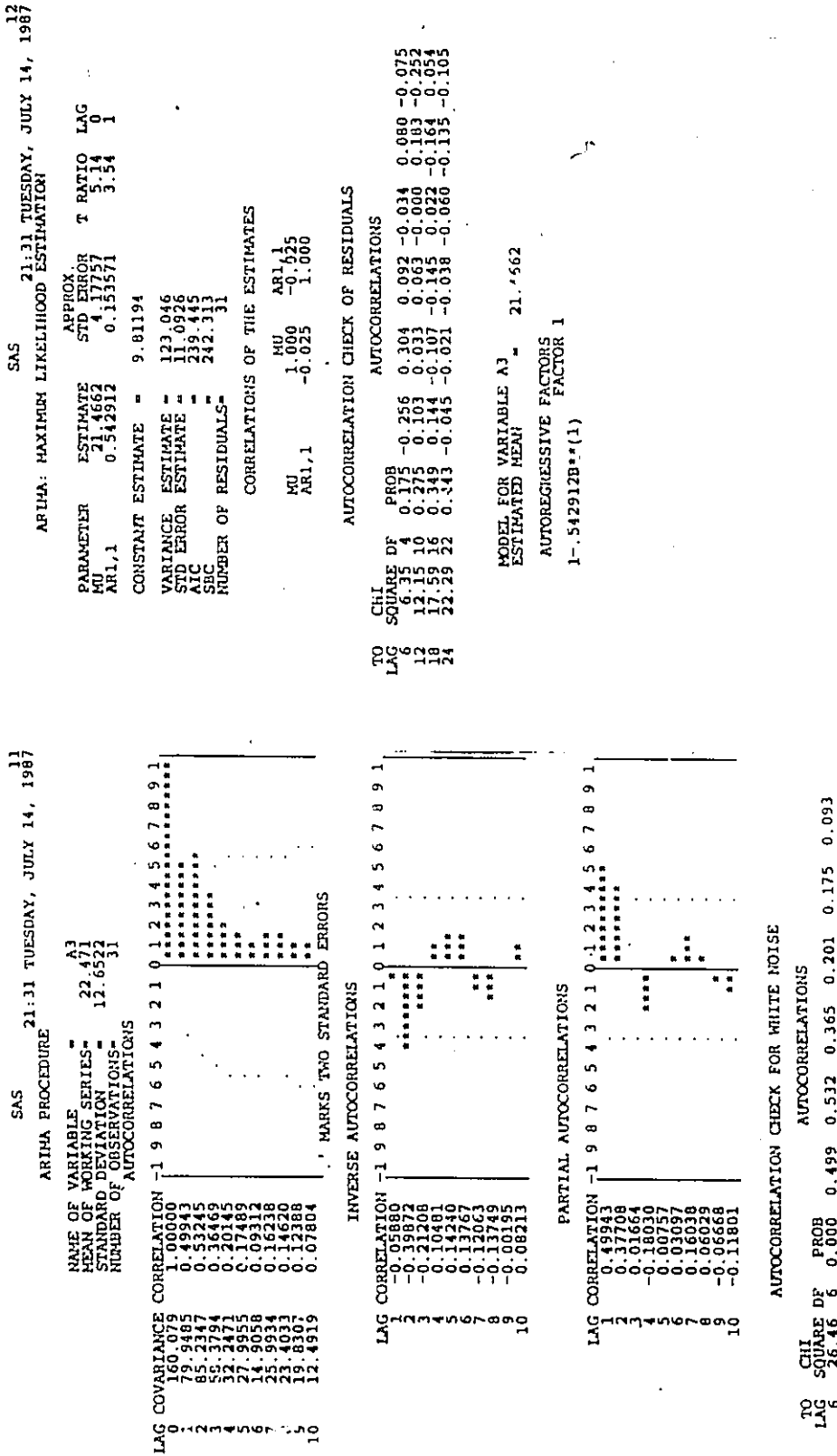
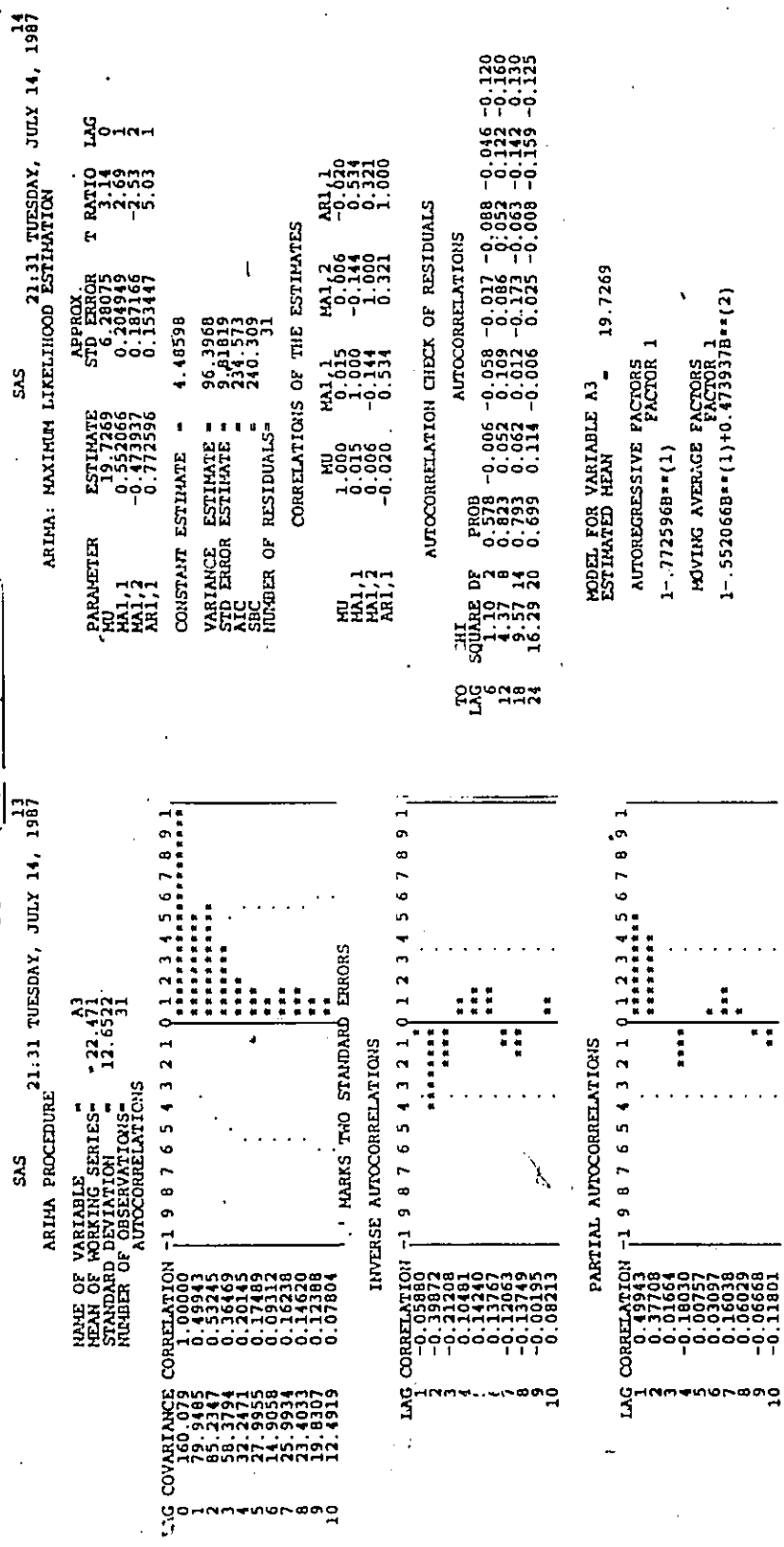


Figure 4.12 Printout of estimated a.c.f.'s and p.a.c.f.'s and p.a.c.f.'s after fitting the ARMA(1,2) time series model to A3(E. mosquensis)



that decay gradually to zero. Fig. 4.11 shows a simple exponential decay pattern with  $\hat{\phi}_1 = 0.543$ . The constant  $(C) = \mu (1 - \hat{\phi}_1)$  gives  $C = 9.812$ . Other ARIMA estimation routines automatically test the hypothesis that the true coefficient is zero. A  $t$ -value to test each coefficient for being different from zero is 3.54, which is larger than the 95% confidence limit of 2.0. The coefficient is therefore statistically significant. The p.a.c.f supports this as a good model for the time series. The  $t$ -value for the p.a.c.f at lag 1 is 2.78 and this value is also greater than 2.0. Consequently  $\hat{\phi}_{11}$  is also different from zero at the 5% significance level.

If the random shock process was tested as the null hypothesis by calculating how many standard errors ( $t_b$ ) away from zero each residual autocorrelation coefficient falls, it follows that  $\hat{a}_1 = -0.256$  with standard error 0.179 and  $t_b$ -value of -1.43. The absolute value of this  $t_b$  value is larger than 1.25 for lag 1, so the residuals are significantly autocorrelated. If the residuals are autocorrelated, they are not white noise. Another model must be found with residuals that are consistent with the independence assumption. Finally, AR(1) is not a good model for this situation.

For the model ARMA(1,2), the test of p.a.c.f gives  $(\hat{\phi}_{kk}) = 0.197$ . The null hypothesis  $H_0: \rho_1 = 0$  results in

$$t_{\hat{\phi}_{11}} = \frac{0.499 - 0}{0.197} = 2.53$$

The absolute value of the  $t$ -statistic greater than 2.0 and  $\hat{\phi}_{11}$  is

different from zero at the 5% significance level. We also have

$$\hat{\phi}_1 = 0.77$$

$$\hat{\theta}_1 = 0.55$$

$$\hat{\theta}_2 = -0.47 \quad (\text{from Fig. 4.12})$$

The stationarity requirement for ARIMA ( 1, 2 ) with  $p = 1$  ,  $q = 2$  is that  $|\hat{\phi}_1| < 1$  for the AR coefficient. Further,

$$\theta_2 + \theta_1 < 1, \quad 0.076 < 1$$

and,

$$\theta_2 - \theta_1 < 1, \quad -1.023 < 1$$

It may be concluded that the A3 ( Epistomina mosquensis ) variable is stationary. Its autocorrelation coefficients decrease gradually to zero.

The random shock ( white noise ) was tested with ARIMA ( 1,2 ). The standard error is 0.179 and the first residual value is - 0.355. Its absolute value is less than 1.25 for lag 1. For lag 2, the  $t_b$  - value is 0.18 and is less than 1.25 too. The random shocks (  $a_t$  ) in model ARMA ( 1, 2 ) are assumed to be statistically independent. The chi - squared test for random shocks shows that the calculated value is 1.10 for  $df = 2$ . The critical value with  $df = 2$  at the 5% significance level is 5.99 ( chi - squared table taken from statistical tables). The calculated chi - squared value is less than the critical value. The conclusion is that the residual autocorrelations are not significantly different from zero. The hypothesis is accepted that the random shocks are independent. The model ARMA ( 1,2 ) provides a good fit for working with the data set. This model becomes:

$$X_t = 4.48 + 0.77X_{t-1} - 0.55a_{t-1} + 0.474a_{t-2} + a_t$$

Now that a model which provides a good fit to the data set of Epistomina mosquensis has been selected, geostatistical modeling can be applied. As shown in Fig. 4.12, the autocorrelation plotted against the lag has an exponential trend ( signal - plus - noise model ). If the vertical axis has a logarithmic scale (as in Fig. 4.13 see page 83 ), the exponential curve becomes a straight line. The stationary series has constant values  $c$  and  $a$  which are estimated from the correlogram. We have

$$a = 0.22 \text{ and}$$

$$c = 0.76$$

Consequently, Eq. 47 becomes

$$r(k) = 0.76 e^{-0.22k}$$

$S^2(x)$  are two or more than two adjacent variances which are calculated with the F FORTRAN program ( Appendix I ).  $F_N$  values are relative variances which are constant. They are calculated according to Eq. 54 as shown in table 4.4. The  $F_L$  values are theoretical relative variances which are calculated using Eq. 59 as seen in Table 4.4.

TABLE 4.4 Theoretical and observed relative variance of Epistomina Mosquensis

$k$	$S^2(x)$	$F_N$	$F_L$
-----	----------	-------	-------

0	0	0	0.24
1	165.42	1	0.95
2	123.42	1.49	1.56
3	108.48	1.97	2.09
4	95.01	2.29	2.55
5	80.60	2.44	2.96
6	69.77	2.53	3.31

The theoretical and observed relative variance values were plotted against lag as shown in Fig. 4.14. The correspondence between experimental and theoretical values is a good fit.

$$\text{The mean standard deviation is } s(\bar{x}_{31}) = \frac{0.1286}{(31)^{1/2}} = 0.023$$

The population mean ( $\mu$ ) lies in the interval  $x \pm 0.023t_{0.975}$  (two tails) with a probability of 95% or because  $t_{0.975}(30) = 2.06$   
 $P(0.177 < \mu < 0.272) = 95\%$ . Using Eq. 54 we obtain,  
 $s(\bar{x}_{31}) = (31)^{1/2} s(x) (F_{31})^{1/2}$ . This means that estimated values above  $(x_{31}) = 0.023$  must be corrected by the factor  $(F_{31})^{1/2}$ . The value  $F_{31}$  can be calculated as 6.14.

$$F_{31} = 6.14 \text{ and } (F_{31})^{1/2} * 0.023 = 0.057$$

The above confidence interval for the mean must be replaced by

$$P(0.168 < \mu < 0.282) = 95\%$$

From Fig. 4.12 the estimated mean is 0.01972 which remains

# THEORETICAL CURVE

A3:E.MOSQUENSIS

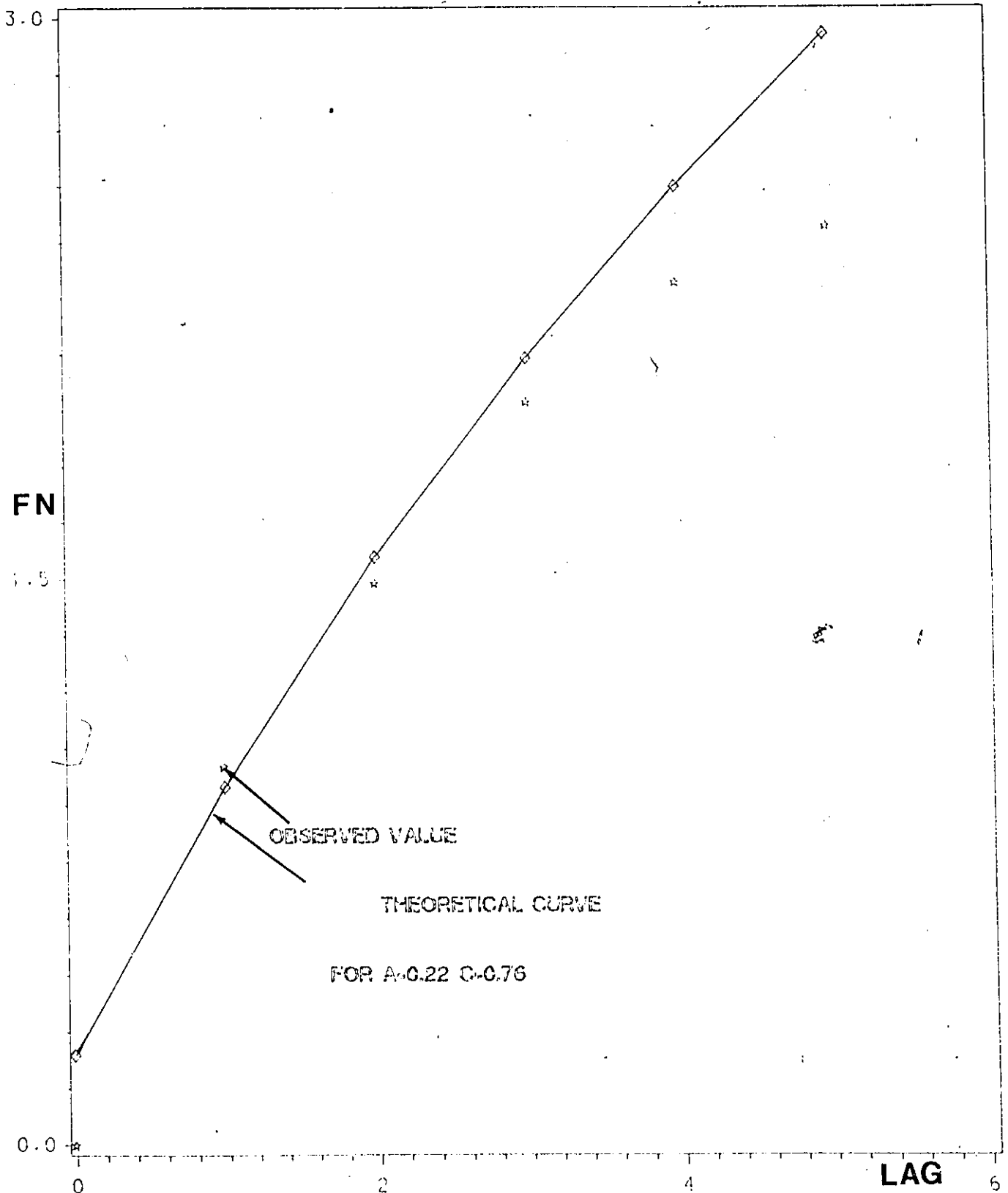


Figure 4.14 Theoretical curve for  $a = 0.22$  and  $c = 0.76$  in comparison with observed values of A3(E. mosquensis)

Table 4.5 Probit transformation of A3 ( Epistomina mosquensis ) from the Tojeira 1 section.

Samples single %	probit	sample pairs %	probit	sample set of 4 %	probit
0.0	2.187				
0.0	2.187	0.0	2.187		
0.0	2.187				
0.2	2.187	0.1	2.187	0.0	2.187
15.1	3.968				
8.4	3.621	11.75	3.812		
24.2	4.300				
18.1	4.081	21.15	4.198	16.45	4.024
12.4	3.845				
26.1	4.360	19.25	4.131		
27.6	4.405				
13.2	3.883	20.4	4.173	19.8	4.151
23.6	4.281				
23.3	4.271	23.45	4.276		
15.5	3.985				
35.2	4.620	25.35	4.336	24.4	4.304
21.4	4.207				
46.3	4.907	33.85	4.583		
18.0	4.085				

( continued )

Table 4.5 (concluded)

39.7	4.739	28.85	4.442	31.35	4.514
30.4	4.487				
13.5	3.897	21.95	4.226		
23.0	4.261				
33.5	4.574	28.25	4.424	25.1	4.329
37.1	4.671				
36.6	4.658	36.81	4.664		
38.8	4.715				
34.8	4.609	36.8	4.663	36.8	4.663
36.5	4.655				
27.4	4.399	31.95	4.531		
16.7	4.034				

Table 4.6 Cumulative frequency distributions of the Epistomina mosquensis for 31, 15 and 7 samples.

N = 31

Class interval	Freq.	Cum freq	Cum %	Plott. value
2.1 - 2.7	4	4	12.90	11.70
2.7 - 3.3	0	4	0.000	11.70
3.3 - 3.9	4	8	12.90	11.70
3.9 - 4.5	14	22	70.97	69.15
4.5 - 5.1	9	31	100.0	97.88

N = 15

2.0 - 2.8	2	2	13.13	10.9
2.8 - 3.6	0	2	13.13	10.9
3.6 - 4.4	7	9	60.0	56.5
4.4 - 5.2	6	15	100.0	95.7

N = 7

2.0 - 2.8	1	1	14.29	9.1
2.8 - 3.6	0	1	14.29	9.1
3.6 - 4.4	6	7	100.0	90.1

within the confidence interval.

The raw data set for single, paired and sets of 4 of adjacent values of Epistomina mosquensis series were transformed with the probit transformation ( Table 4.5 ) using table. The frequency distribution for these data is shown in Table 4.6. Calculated cumulative frequency distributions were plotted on normal probability paper using upper class limits. They are approximately normal distributions which plot as straight lines. This distribution has a mean  $\bar{x} = 0.2247$  and a variance  $s^2(x) = 0.016538$  for 31 values of A3. The R can be obtained from the mean and variance relationships. However, the mean and variance were too large to be found on standard prepared tables and graphs. Because of this, the R values were estimated from a different graph which has been plotted by Agterberg (1984).

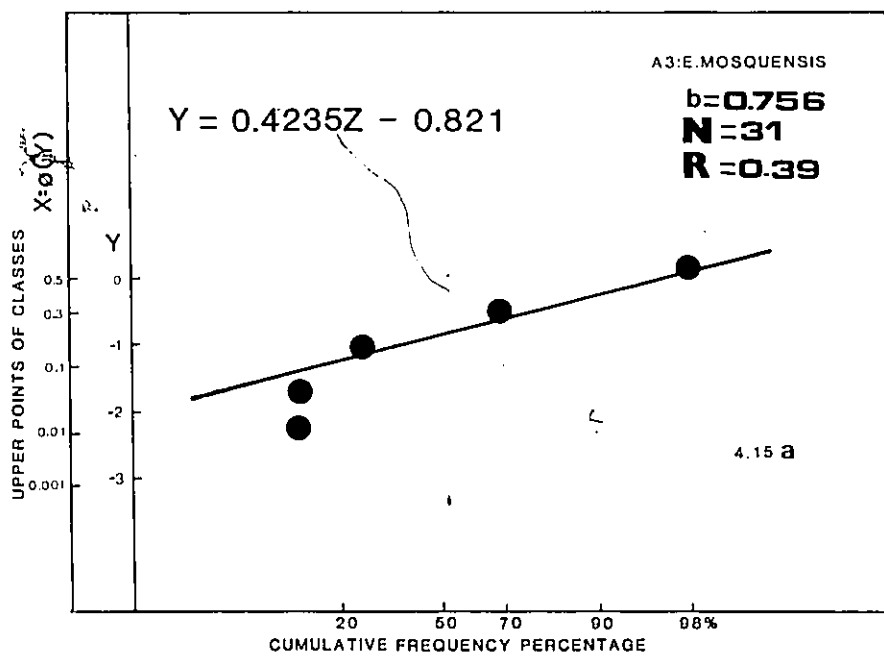
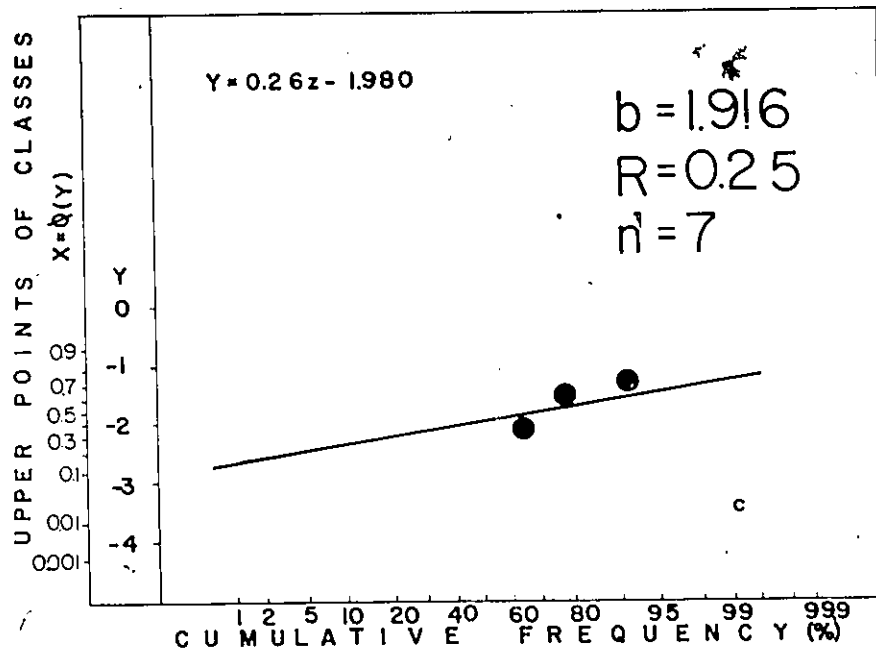
The scale of this graph is so small that R values estimated from this graph are not very precise. The estimated R value is 0.39.

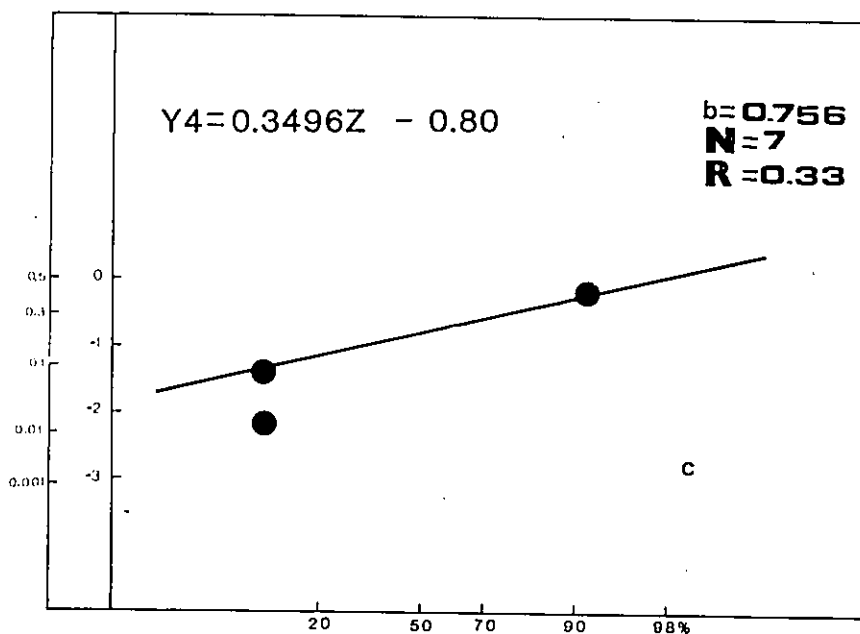
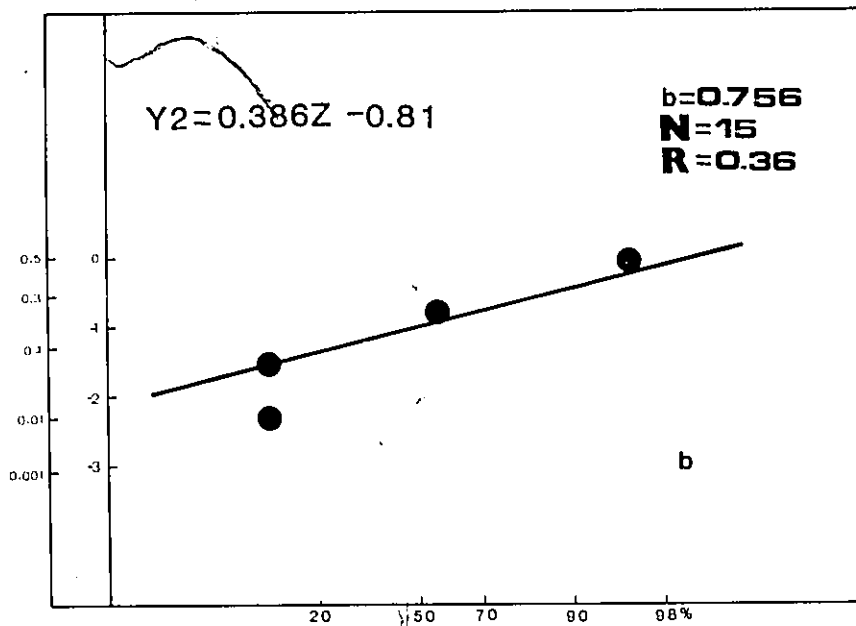
For Eq. 62 the coefficient  $\Phi(b) = 1 - m = 1 - 0.2247 = 0.7753$ . The probit value in Table is 5.756 for  $\Phi = 0.775$ . Calculating  $b = 5.756 - 5 = 0.756$ . Eq. 60 becomes,  $Y = 0.4235Z - 0.821$ .

This function is plotted on normal probability paper ( Fig 4.15a ), which is the same paper used for the probit transformed single data of Epistomina mosquensis, using the values shown below.

Z = 0	y = - 0.821
Z = 1	y = - 0.397
Z = 2	y = 0.025

Figure 4.15a, b, c Experimental frequency and theoretical (probnormal) distributions for 31 (single values) 15 (pairs) and 7 (sets of four adjacent samples) for A3(E. mosquensis). The two distributions show a good fit.





$$Z = -1 \quad y = -1.244$$

$$Z = -2 \quad y = -1.668$$

The distribution of the 15 successive values ( $31/2 = 15$ , pairs) of A3 has the same mean as that of the single values of A3. Its variance is

$$S^2(x_2) = \frac{1.49 * 0.01654}{2} = 0.01232$$

For this mean and variance, the R value estimated from Agterberg's (1984) graph is 0.36. The value b is constant and it is dependent on the mean. Eq. 60 becomes  $Y_2 = 0.386z - 0.81$  for these values. The  $Y_2$  function is also plotted on the probability paper and shown in Fig. 4.15b. The 7 successive relative values of A3 ( $15/2 = 7$  or  $31/4 = 7$ , sets of four of samples) also have the same mean. Its variance is

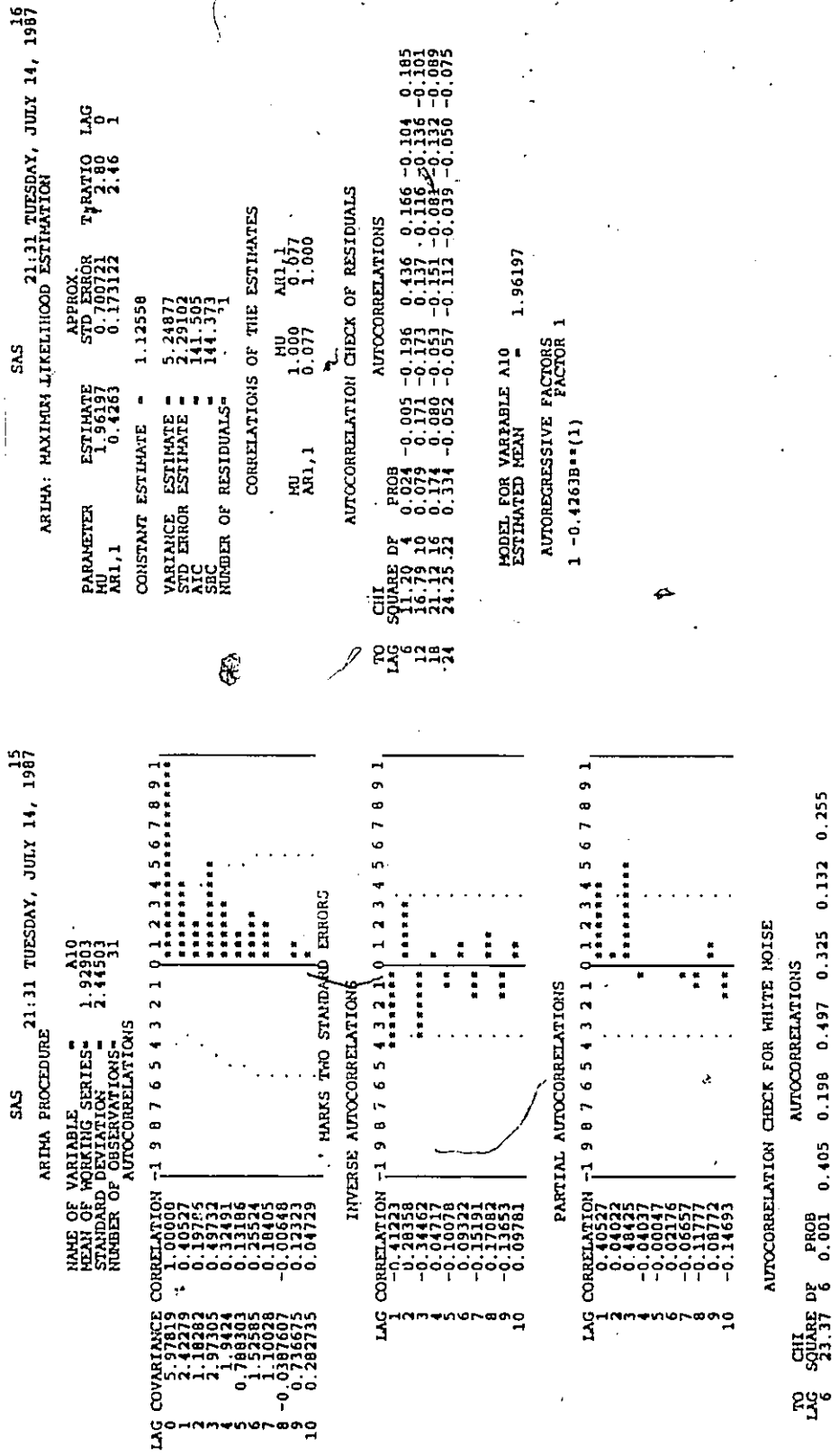
$$S^2(x_4) = \frac{2.29 * 0.01654}{4}$$

and by using Agterberg's graph,  $R = 0.33$ . The  $Y_4 = 0.3496z - 0.80$  function is plotted as the others and shown in Fig. 4.15c. There is good agreement between the observed values and the theoretical probability normal distribution as shown in Fig. 4.15a, b and c.,.

#### 4.2.3 OPHTHALMIDIUM STRUMOSUM

The ARIMA procedure was applied to A10. The model AR(1) and

Figure 4.16 Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to A10(C. strumosum)



ARMA (1,1) were fitted to the data set of Ophthalmidium strumosum AR(1) and ARMA (1, 1) appear to be good models for this series. Their first and third autocorrelations are significantly different from zero as seen in Fig. 4.16. Here we see the first and second asterisks in the a.c.f extend beyond the 95% confidence belt of two standard errors represented by dots. The AR (1) and ARMA (1, 1) are stationary since the absolute value of  $\hat{\phi}_1 < 1$ . The key point to remember is that any stationary AR (1) process has a theoretical a.c.f showing exponential decay. ARMA (1, 1) model also shows exponential decay when  $\hat{\phi}_1$ 's are positive.

The quality of estimated coefficients can be tested with the t- test. For AR (1),  $t_1 = 2.46$  ( see Fig.4.16 ) which is greater than 2.0 and  $\hat{\phi}_1$  is significantly different from zero at the 5% significance level. For ARMA(1, 1),  $t_{\phi_1} = 6.16$ ,  $t_{\theta_1} = 2.58$  and  $\hat{\phi}_1$  and  $\hat{\theta}_1$  are larger than 2.0. This means that the coefficients of both models are good estimates ( see Fig. 4.17).

The  $t_b$  - test was used for random shocks, using Eq. 23. For AR (1),  $t_{b_1} = -0.279$  at lag 1 and  $t_{b_2} = 0.945$  at lag 2. The absolute  $t_b$  values of the residual a.c.f are less than 1.25 at lags 1 and 2. The random shocks at those lags are independent and not correlated. For ARMA (1, 1),  $t_{b_1} = 0.173$  at lag 1 and  $t_{b_2} = 0.533$  at lag 2. The absolute  $t_b$  values are less than 1.25, indicating that the random shocks are not correlated for the ARMA (1, 1) model either.

Furthermore, the chi - squared test was applied to the two models. For AR (1), the calculated value is 11.20 for degrees of freedom



df = 4. The critical value from table is 9.49 for df = 4. The calculated value is 16.79 for df = 10. The critical value is 18.3 for df = 10 at the 5% significant level. Since the calculated chi - squared value is less than the critical values it may be concluded that the random shocks are independent ( Fig. 4.16 ).

ARMA ( 1,1 ) has a calculated chi - squared value of 8.10 at lag 6 for df = 3 and a calculated value of 12.30 at lag 12 for df = 9. The critical values are 7.81 and 16.9, respectively, at the 5% significance level. The calculated chi - squared value is less than the critical value at lag 6, so the chi - squared statistics suggest that the ARMA ( 1, 1 ) model is inadequate ( Fig. 4.17 ).

If AR ( 1 ) and ARMA ( 1, 1 ) were good models for the working series, AR ( 1 ) would be preferred because it has one estimated value fewer than model ARMA ( 1, 1 ).

A10 ( Ophthalmidium strumosum ) has a time series model which is AR ( 1 ). AR ( 1 ) can be written as  $X_t = C + \phi_1 X_{t-1} + a_t$ . This model is a signal plus - noise model.

The geostatistical model can be applied as to the previous data set ( Epistomina mosquensis ). The correlation coefficients were plotted against the lag. Its pattern is an exponential curve. When a logarithmic scale is used for the correlation coefficients, as in Fig. 4.18, this exponential curve plots as a straight line. The a and c values are constant for a stationary series and are estimated from Fig. 4.18. The values are a = 0.08 and c = 0.50. Eq. 47 becomes

$$r(k) = 0.50e^{-0.08k}$$



$F_N$  and  $F_L$  were calculated as in section 4.21 and Table 4.7 was constructed.

TABLE 4.7 Theoretical and observed relative variance of Ophthalmidium strumosum

k	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.50
1	6.18	1	0.99
2	4.36	1.41	1.45
3	3.54	1.72	1.89
4	3.33	2.16	2.30
5	3.16	2.56	2.70
6	2.96	2.87	3.07

The observed relative variance ( $F_N$ ) and theoretical variance ( $F_L$ ) were plotted against lag. The correspondence between values for  $F_N$  and  $F_L$  is shown in Fig. 4.19. For large values of N or L the curve tends towards a horizontal limit of  $F_N = 8.38$ . The standard deviation of 31 uncorrelated data would be  $s(\bar{x}_{31}) = 0.004$ . The population mean ( $\mu$ ) lies in the interval  $\bar{x} \pm 0.004 * t_{0.975}$  with a probability of 95% because  $P(0.0102 < \mu < 0.028) = 95\%$  According to Eq. 59,  $F_{31} = 8.38$ ,  $s(x_{31})$  must be corrected by the factor  $(F_{31})^{1/2} * s(x_{31}) = (8.38)^{1/2} * 0.004 = 0.12$ . This results in a new confidence interval of  $P(0.008 < \mu < 0.0309) = 95\%$ . The estimated mean in Fig. 4.16 is 0.0196 which falls within this confidence belt.

The Ophthalmidium strumosum data, pairs of Ophthalmidium

## THEORETICAL CURVE

A10:O.STRUMOSUM:A

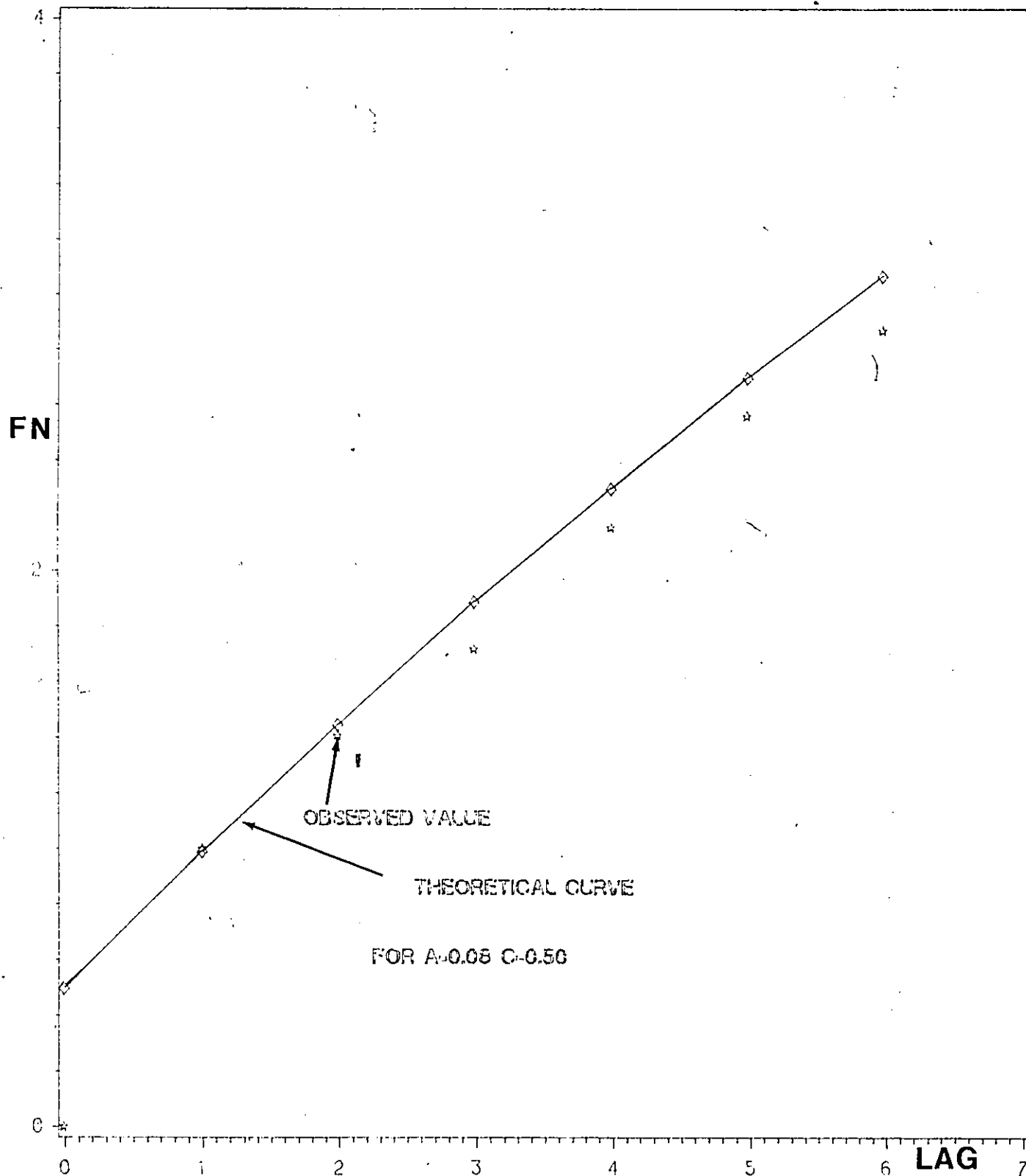


Figure 4.19 Theoretical curve for  $a = 0.08$  and  $c = 0.50$  in comparison with observed values of O. strumosum

Table 4.8 probit transformation of A10 (Ophthalmidium strumosum)  
from the Tojeira 1 section.

sample single %	probit	sample pairs %	probit	sample set of 4 %	probit
0.0	2.187				
0.0	2.187	0.0	2.187		
0.0	2.187				
0.0	2.187	0.0	2.187	0.0	2.187
0.0	2.187				
0.0	2.187	0.0	2.187		
0.0	2.187				
0.2	2.122	0.1	1.910	0.05	2.187
0.0	2.187				
0.0	2.187	0.0	2.187		
0.3	2.252				
0.2	2.122	0.25	2.187	0.12	1.910
0.0	2.187				
0.0	2.187	0.0	2.187		
2.3	3.005				
0.3	2.252	1.3	2.774	0.65	2.510
2.9	2.634				
8.3	3.615	5.6	3.411		
6.6	3.494				
1.9	2.925	4.2	3.279	4.91	3.345

( continued )

Table 4.8 ( concluded )

0.4	2.348				
7.2	3.539	3.8	3.226		
2.4	3.023				
1.9	2.925	2.15	2.976	2.96	3.112
5.8	3.428				
1.9	2.925	3.85	3.232		
2.9	3.104				
5.2	3.374	4.05	3.255	3.95	3.243
2.7	3.073				
1.1	2.710	1.90	2.925		
5.3	3.384				

Table 4.9 Cumulative frequency distributions of A10 (Ophthalmidium strumosum) for 31, 15 and 7 samples.

N = 31

class interval	freq %	cum freq	cum %	plott. value
2.25 - 2.55	16	16	51.61	50.0
2.55 - 2.85	2	18	58.06	56.4
2.85 - 3.15	7	25	80.65	78.7
3.15 - 3.45	3	28	90.32	88.3
3.45 - 3.75	3	31	100.0	97.9

N = 15

1.8 - 2.2	7	7	46.67	43.5
2.2 - 2.6	0	7	46.67	43.5
2.6 - 3.0	3	10	66.67	63.0
3.0 - 3.4	5	15	100.0	95.7

N = 7

1.75 - 2.25	3	3	42.86	36.4
2.25 - 2.75	1	4	57.14	50.0
2.75 - 3.25	3	7	100.0	90.9

strumosum data and set of 4 adjacent Ophthalmidium strumosum data were transformed with the probit transformation ( Table 4.8). The frequency distribution table for these data was constructed as Tables 4.9. Their cumulative frequency distributions were plotted on normal probability paper. The mean of the 31 observed values of A10 (Ophthalmidium strumosum ) is,

$$\bar{x} = 0.01929$$

its variance is,

$$s^2(x) = 0.000598 \text{ and}$$

$b = 2.068$  which is constant.

$R = 0.43$  and was derived from the mean and variance, using Appendix II or III. Eq. 60 becomes  $Y = 0.4762Z - 2.2290$  which is the prob - normal theoretical distribution model for  $N = 31$ .

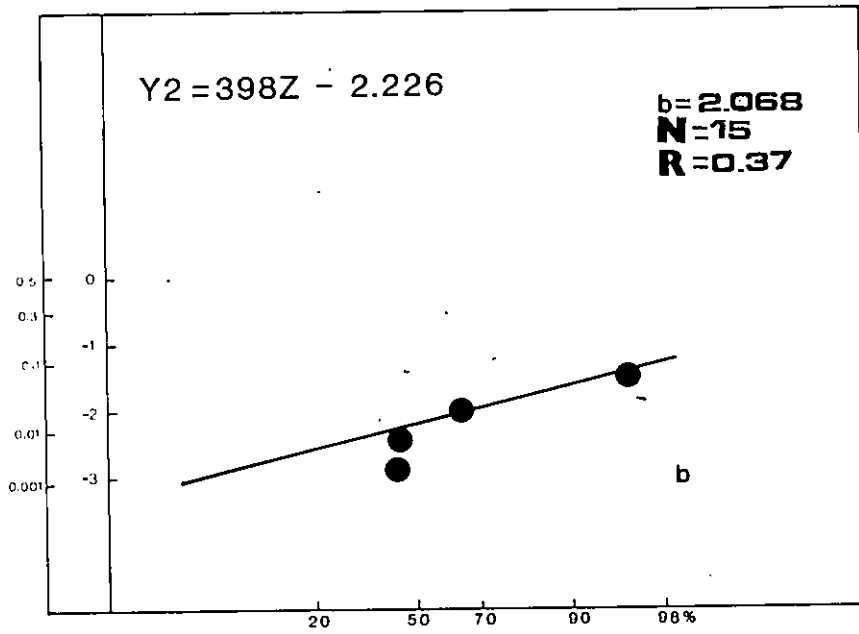
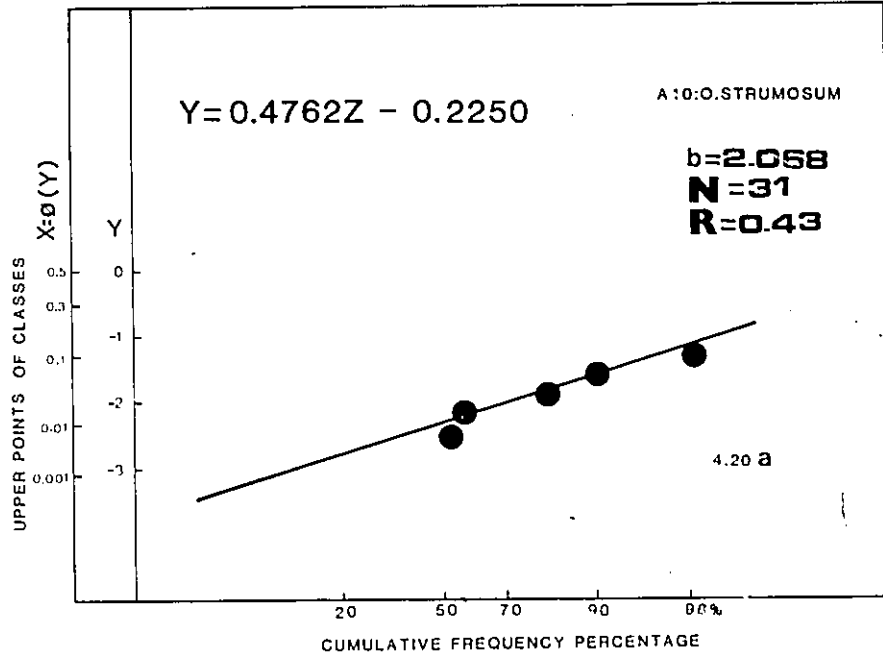
The theoretical distribution of 15 pairs of successive values of A10, shows that the mean is the same,  $\bar{x} = 0.01929$  the variance  $s^2(x) = 0.0004216$  and  $b = 2.068$   $R = 0.37$  and again was derived from the mean and variance using Appendix II or III. The theoretical distribution function is  $Y_2 = 0.398Z - 2.2206$

The theoretical distribution of 7 means of four successive relative values of A10 is  $Y_4 = 0.349Z - 2.190$  for the mean  $\bar{x} = 0.01929$  variance,  $s^2(x) = 0.0003229$  and  $R = 0.33$   $b = 2.068$

They were plotted on probability paper as can be seen in Fig. 4.20a, b, and c and determine the positions and slopes of the lines for prob - normal distributions satisfying Eq. 60.

4.3 - Tojeira 2 section

Figure 4.20a,b,c Experimental frequency distributions for 31,15 and 7 values of  $A_{10}(Q. \text{strumosum})$  in comparison with their theoretical distributions (probnormal distributions) satisfying Eq.60



#### 4.3.1 Eoguttulina sp., Epistomina mosquensis, Ophthalmidium strumosum, Spirillina tenuissima and agglutinants

An a.c.f and p.a.c.f were estimated from the data set of Tojeira 2 (see Appendix I). Printout showed that 5 of 14 taxa had theoretical autocorrelations statistically different from zero at lag 1, 2 or 3. The 5 taxa are B1, B2, B11, B13, and B14. Their mean of these realizations is probably stationary. The estimated autocorrelations drop to zero fairly quickly with approximate exponential decay. Stationarity is confirmed since;

for taxon B1 (Eoguttulina sp.)  $\hat{\phi}_1 = 0.475$ ,

for taxon B2 (Epistomina mosquensis)  $\hat{\phi}_1 = 0.462$ ,

for taxon B11 (Spirillina tenuissima)  $\hat{\phi}_1 = 0.748$ ,

for taxon B13 (Ophthalmidium strumosum)  $\hat{\phi}_1 = 0.805$  and

for taxon B14 (agglutinants)  $\hat{\phi}_1 = 0.51$

These satisfy the condition  $|\hat{\phi}_1| < 1$ . The estimated coefficients are acceptable because the corresponding absolute t - values are equal to or greater than 2.0. All are significantly different from zero at the 5% significance level. However, the t- value of the  $\hat{\phi}_2$  for taxon B2 is less than 2.0. In fact, it can be assumed to be equal to 2.0

( $\hat{\phi}_2 = 1.97$ )

The purpose of having a good time series model is to compare the estimated a.c.f and p.a.c.f with various theoretical a.c.f.'s and p.a.c.f.'s to find a match. The various theoretical a.c.f. models have been used to find best fitting models for the data sets of B1, B2, B11, B13, and B14 respectively.

Figure 4.21a Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to B14(Agglutinants).

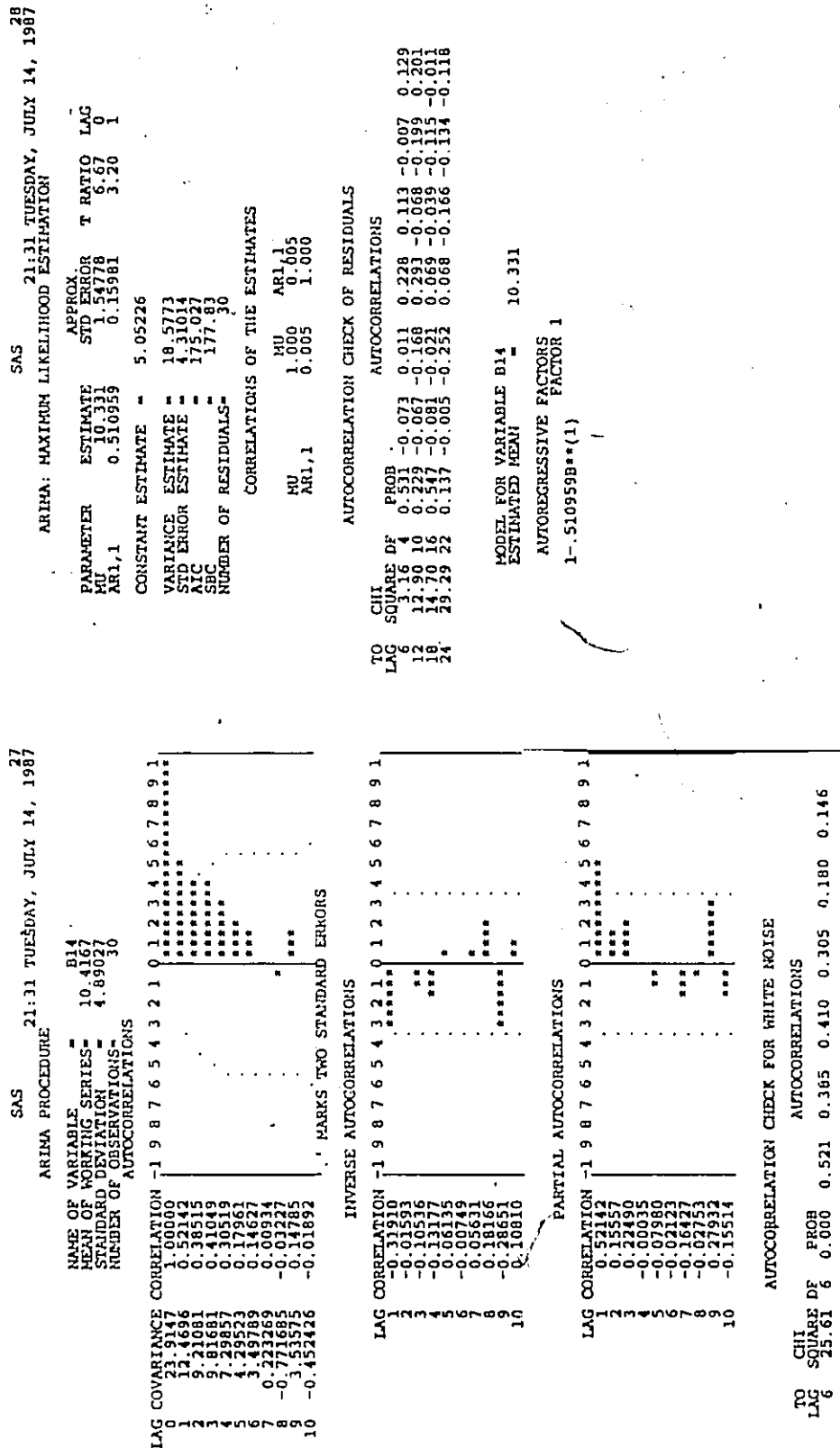
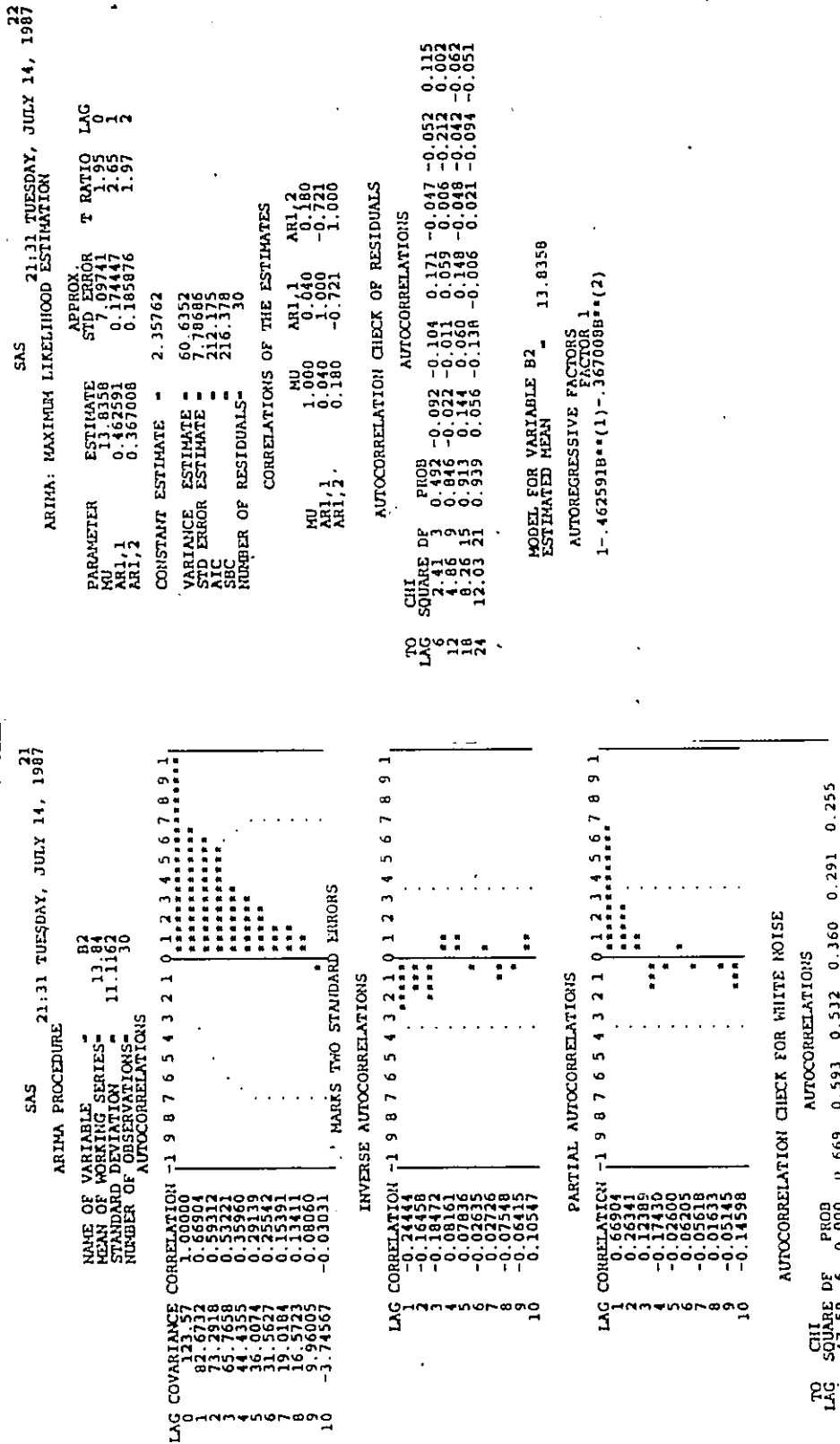


Figure 4.21b Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to B2(E. mosquensis)



Using the tests that were explained before, it was decided that AR ( 1 ) is good model for B14 ( Fig. 4.21a ), and AR( 2 ) is a good model for taxon B2 ( Fig. 4.21b ).

According to these tests, AR ( 1 ) would also be the best model for taxa B1, B11 and B13 ( Figs. 4.21c, 4.21d and 4.21e ). However, the calculated residual of B1 is not statistically zero at lag 2 as seen below.

$$t_{b_1} = \frac{-0.320}{0.1833} = -1.74$$

The absolute  $t_b$  value is larger than 1.25 at lag 2. This suggests that the AR(1) model is not an adequate model. Because the residuals are significantly correlated at lag 2, the residuals absolute  $t_b$  value of B2 is

$$t_{b_2} = \frac{-0.242}{0.1826} = -1.32$$

at lag 1 ( Eq.33 ). Consequently, the random shock is correlated at lag 1 for taxon B11.

Furthermore, the calculated chi - squared value for taxa B11 is large. This suggests that the residual autocorrelations as a set are significantly different from zero, and the random shocks of the estimated model are probably autocorrelated.

In this case, the residual a.c.f. at lag 2 for taxon B1 and lag 1 for B11 might be ignored, believing that one significant autocorrelation out of 8 could occur just because of sampling error. Alternatively, these models could be assumed statistically inadequate, because some residual a.c.f.  $t_b$  - values exceed the suggested critical values.

However, there is no guarantee, of course, that a better model will be discovered.

A statistically adequate model is one whose random shocks are statistically independent, meaning that they are not autocorrelated ( Autocorrelation coefficients are all statistically zero ). We can not expect all residual autocorrelations to be exactly zero, even for a properly constructed model. The reason is, that the residuals are calculated from a realization ( not a process ) using only estimates of the ARIMA coefficients ( not their true values ). Therefore, we expect that sampling error may cause some residual autocorrelations to be nonzero even if a good model has been found.

Unfortunately, there is a potential problem in using Bartlett's formula ( Eq.33 ) in testing residual autocorrelation. The estimated standard errors are sometimes seriously overstated when applying Bartlett's formula to residual autocorrelations. This is especially possible at the very short lags. If the estimated standard errors are overstated, then the corresponding  $t_b$  - values are underestimated. This is very likely, since finding the exact values for the estimated standard errors and  $t$  - values for residual autocorrelations is relatively difficult ( Box and Pierce, 1970 ).

According to the  $t$  - test ( at lag 2 ) and chi - squared test ( for  $df = 4$  and  $df = 10$  ), the random shocks of the estimated model of taxon B13 are autocorrelated. The various models tested, which are AR ( 1 ), AR ( 2 ), ARMA ( 1, 1 ), ARMA ( 2,1 ), and ARMA ( 1,2 ), are not good time series models for taxon B13. However, the AR ( 1 ) model

Figure 4.2(c) Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to BI(Eoguttulina sp.)

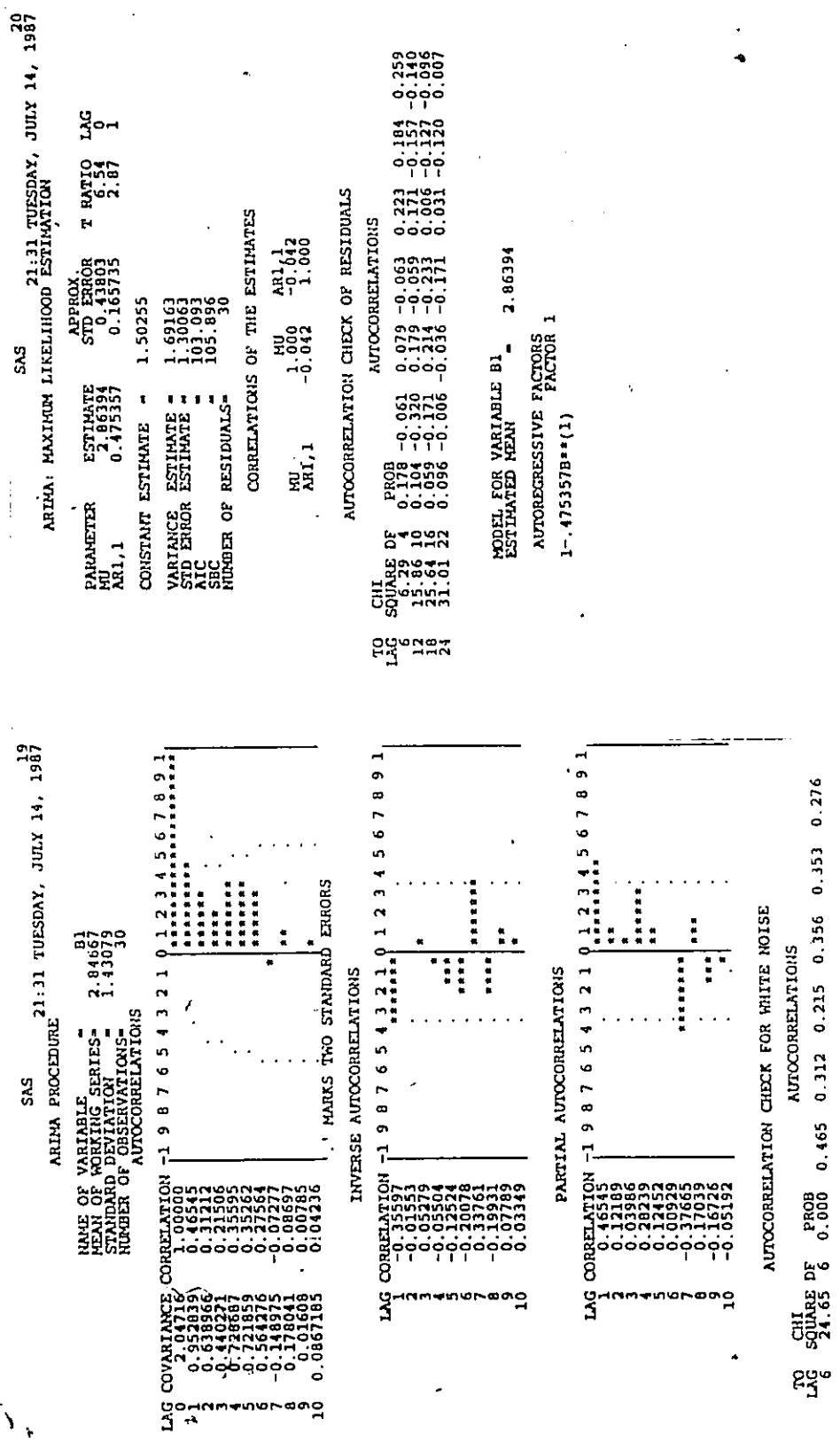


Figure 4.14d Printout of estimated a.c.f.'s and p.a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to Bll(S. tenuissima)

SAS 21:31 TUESDAY, JULY 14, 1987

SAS 21:31 TUESDAY, JULY 14, 1987

ARIMA PROCEDURE  
 NAME OF VARIABLE = Bll  
 MEAN OF MOVING SERIES = 25.7533  
 STANDARD DEVIATION = 17.3525  
 NUMBER OF OBSERVATIONS = 30  
 NUMBER OF AUTOCORRELATIONS = 30

ARIMA: MAXIMUM LIKELIHOOD ESTIMATION  
 APPROX.  
 PARAMETER ESTIMATE STD ERROR T RATIO LAG  
 AR1,1 0.747753 0.121978 6.13 1  
 CONSTANT ESTIMATE = 6.90635  
 VARIANCE ESTIMATE = 152.811  
 STD ERROR ESTIMATE = 12.3617  
 AIC = 238.762  
 SBC = 241.564  
 NUMBER OF RESIDUALS = 30

CORRELATIONS OF THE ESTIMATES  
 MU AR1,1  
 AR1,1 1.000  
 -0.091 1.000

MARKS TWO STANDARD ERRORS  
 LAG COVARIANCE CORRELATION  
 0 301.63 1 0.0000  
 1 282.92 0.61885  
 2 192.118 0.58314  
 3 172.343 0.42391  
 4 142.915 0.47869  
 5 91.4682 0.30769  
 6 58.3148 0.19367  
 7 28.4353 0.09443  
 8 1.85364 0.00549  
 9 -5.76119 -0.01913

AUTOCORRELATION CHECK OF RESIDUALS  
 TO CHI SQUARE DF PROB  
 LAG 6 11.24 4 0.074  
 12 13.10 10 0.238  
 18 14.05 16 0.337  
 24 15.97 22 0.817

INVERSE AUTOCORRELATIONS  
 LAG CORRELATION  
 1 -0.39549  
 2 -0.13276  
 3 -0.31940  
 4 0.31233  
 5 -0.36177  
 6 -0.09598  
 7 -0.07368  
 8 -0.16188  
 9 -0.05475  
 10 0.02755

AUTOCORRELATIONS  
 TO CHI SQUARE DF PROB  
 LAG 6 11.24 4 0.074  
 12 13.10 10 0.238  
 18 14.05 16 0.337  
 24 15.97 22 0.817

PARTIAL AUTOCORRELATIONS  
 LAG CORRELATION  
 1 0.56970  
 2 0.38280  
 3 -0.15411  
 4 -0.30866  
 5 -0.23593  
 6 -0.12503  
 7 -0.22517  
 8 -0.04611  
 9 -0.04606  
 10 0.04606

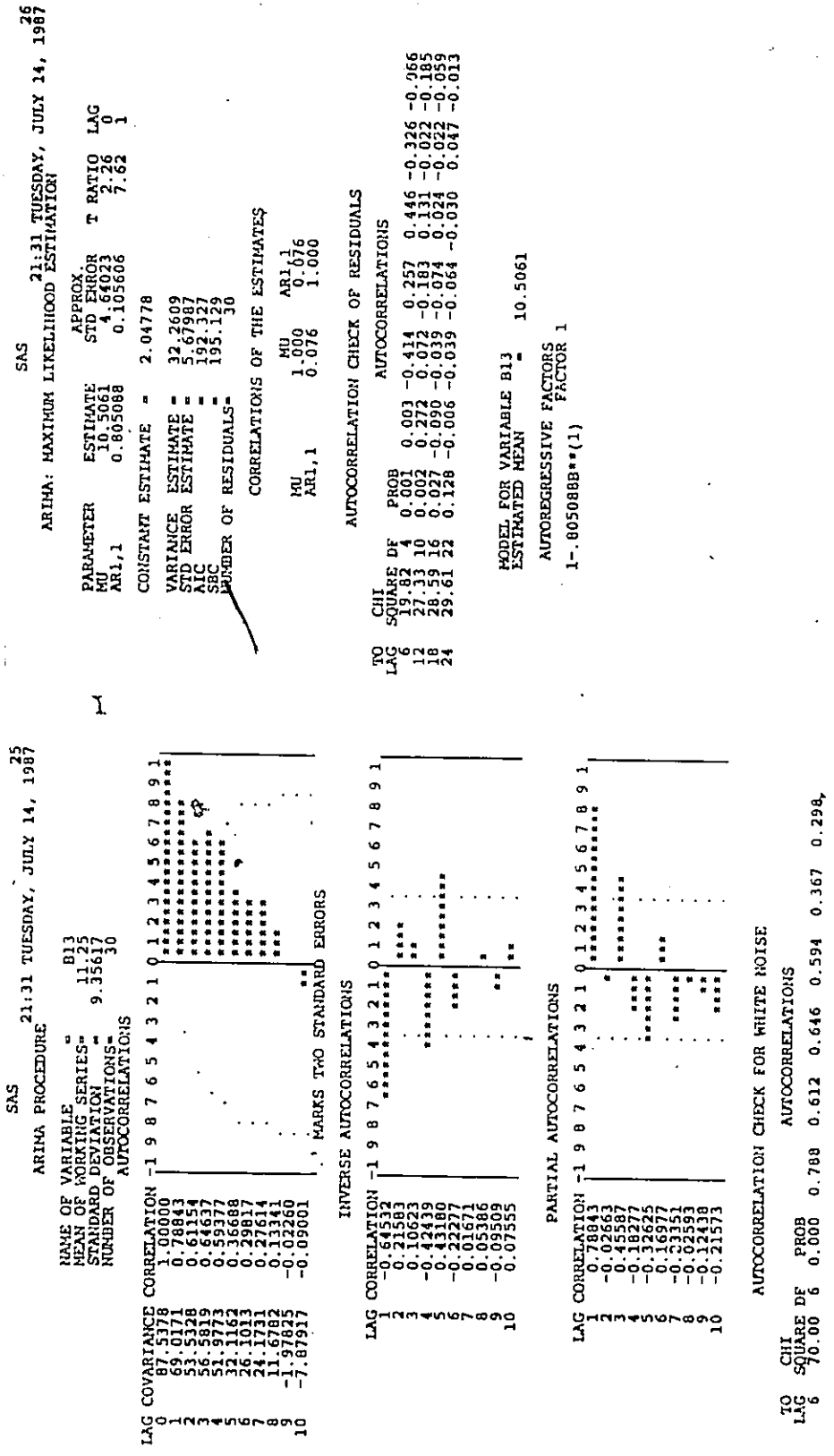
AUTOCORRELATION CHECK OF RESIDUALS  
 TO CHI SQUARE DF PROB  
 LAG 6 11.24 4 0.074  
 12 13.10 10 0.238  
 18 14.05 16 0.337  
 24 15.97 22 0.817

AUTOCORRELATION CHECK FOR WHITE NOISE  
 TO CHI SQUARE DF PROB  
 LAG 6 59.22 6 0.000  
 12 60.00 6 0.677  
 18 61.15 6 0.582  
 24 62.423 6 0.475  
 30 63.307 6 0.307

MODEL FOR VARIABLE Bll = 27.3794  
 ESTIMATED MEAN =  
 AUTOREGRESSIVE FACTORS  
 FACTOR 1  
 1 - .747753B\*\*(1)

AUTOCORRELATIONS  
 TO CHI SQUARE DF PROB  
 LAG 6 11.24 4 0.074  
 12 13.10 10 0.238  
 18 14.05 16 0.337  
 24 15.97 22 0.817

Figure 4.21e Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to B13(O. strumosum)



was selected as a good model ( Fig. 4.21e ) because it has fewer estimated coefficients than the others. A geostatistical model can be applied to the Tojeira 2 section after a time series model has been recognized for each taxon.

The autocorrelation coefficients were plotted against the lag on semi - logarithmic graph paper. The exponential trends become straight lines as can be seen in Fig 4.21. The best fitting lines were drawn and the constant  $c$  and  $a$  values can be obtained as shown in Table 4.10.

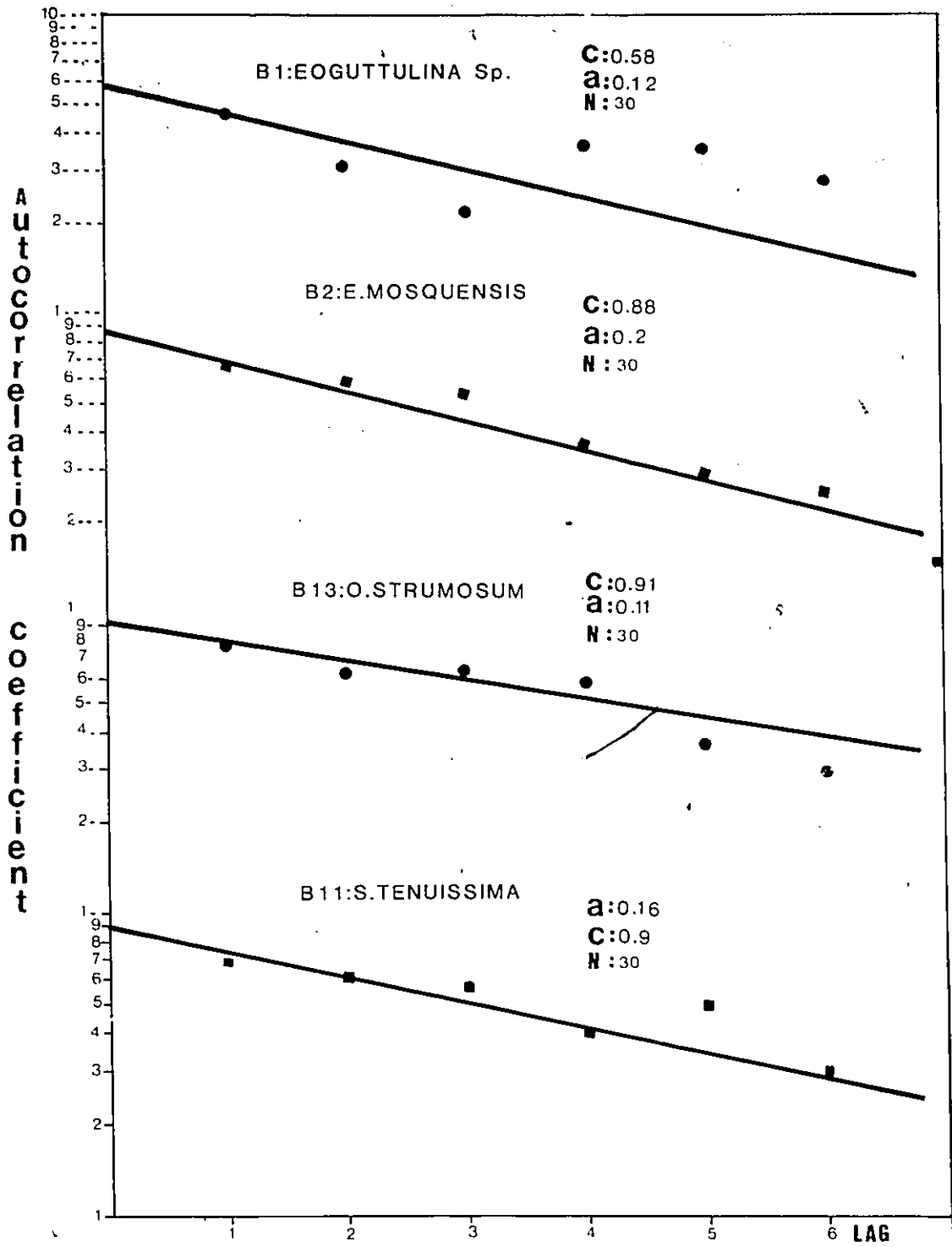
TABLE 4.10 Estimated  $a$ 's and  $c$ 's values from best - fitting lines for B1 ( Eoguttulina sp. ), B2 ( Epistomina mosquensis ), B11 ( Spirillina tenuissima ), B13 ( Ophthalmidium strumosum ) and B14 ( agglutinants )

TAXON	B1	B2	B11	B13	B14
$c$	0.58	0.88	0.90	0.91	0.58
$a$	0.12	0.20	0.16	0.11	0.12

Table 4.10 shows that the  $c$  values for B2, B11 and B13 are quite large or close to 1.0, meaning that these data could fit a time series model which is just a signal ( perfect autocorrelation ) and not a signal - plus - noise model.

Variances of two or more than two adjacent values were calculated with the F FORTRAN program for each taxon  $F_N$  values, which are experimental relative variances, were calculated from variances of adjacent values ( Eq. 54 ).  $F_L$  values which are theoretical relative

Figure 4.21 Correlograms of B1( Eoguttulina sp.), B2(E. mosquensis ), B11(S. tenuissima), B13(O. strumosum) and B14(Agglutinants) with their best - fitting exponential curves. The a's and c's were estimated from the correlograms



# THEORETICAL CURVE

BTEOGUTTULINA SP.

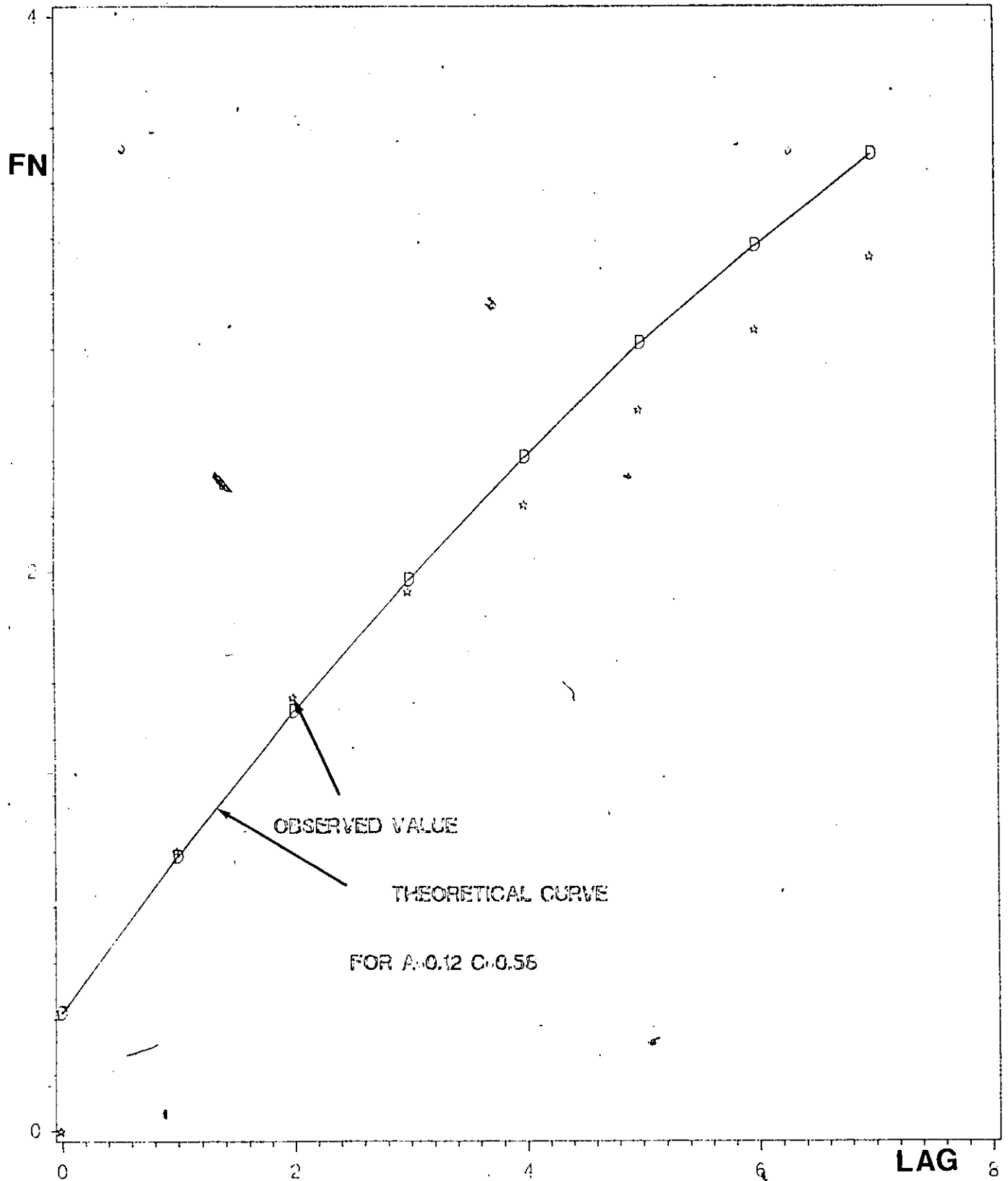


Figure 4.21a<sub>1</sub> Theoretical curve for  $a = 0.12$  and  $c = 0.58$  in comparison with observed values of Bl(Eoguttulina sp.)

# THEORETICAL CURVE

B2.E.MOSQUENSIS

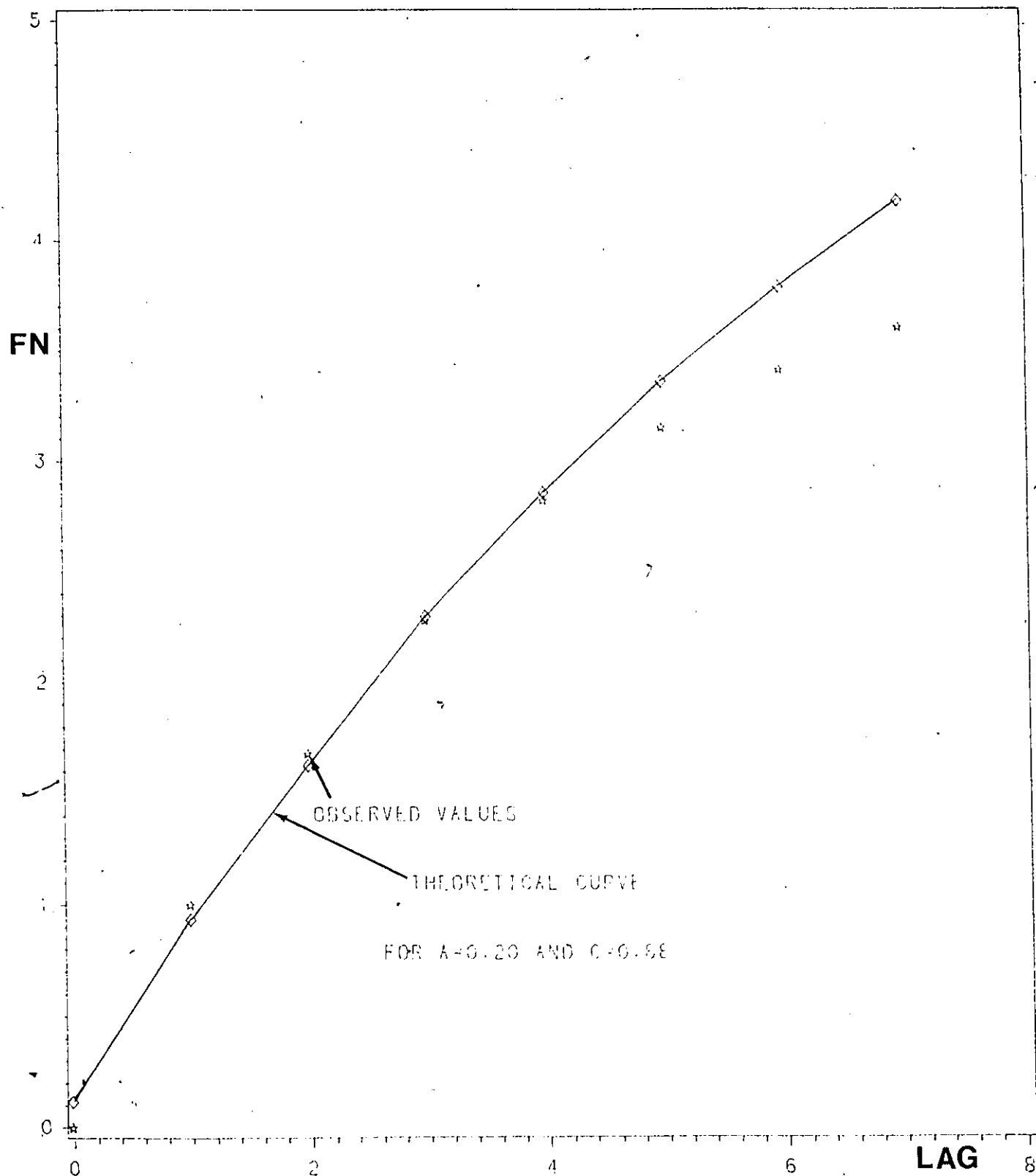


Figure 4.2lb<sub>1</sub> Theoretical curve for  $a = 0.20$  and  $c = 0.88$  in comparison with observed values of  $B2(E. mosquensis)$

## THEORETICAL CURVE

B11(S.TENUISSIMA)

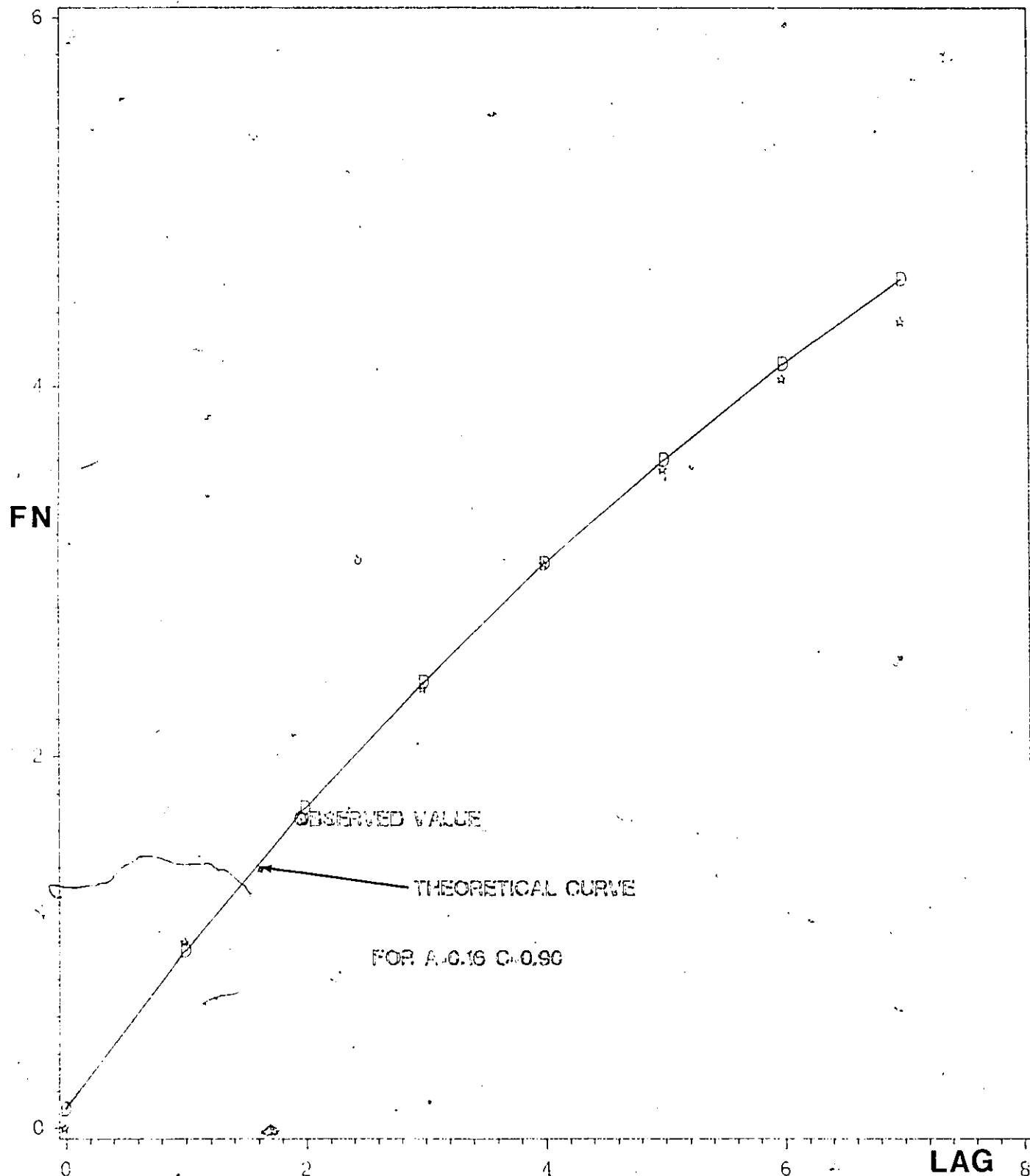


Figure 4.21c<sub>1</sub>. Theoretical curve for  $a = 0.16$  and  $c = 0.90$  in comparison with observed values of  $B_{11}(S. tenuissima)$

# THEORETICAL CURVE

B13:O.STRUMOSUM

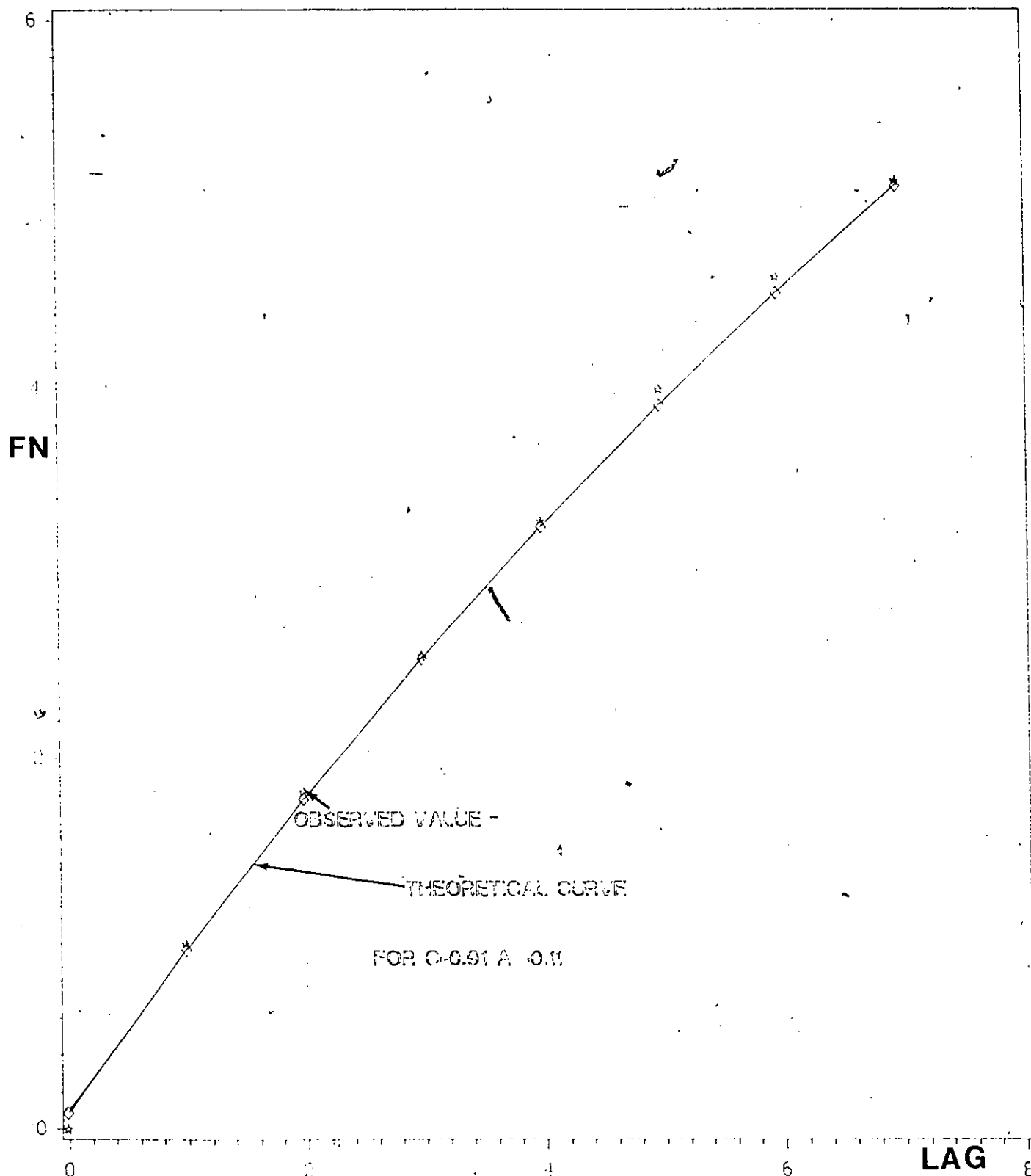


Figure 4.21d<sub>1</sub> Theoretical curve for  $a = 0.11$  and  $c = 0.91$  in comparison with observed values of B13(O. strumosum)

# THEORETICAL CURVE

B14:AGGLUTINANTS

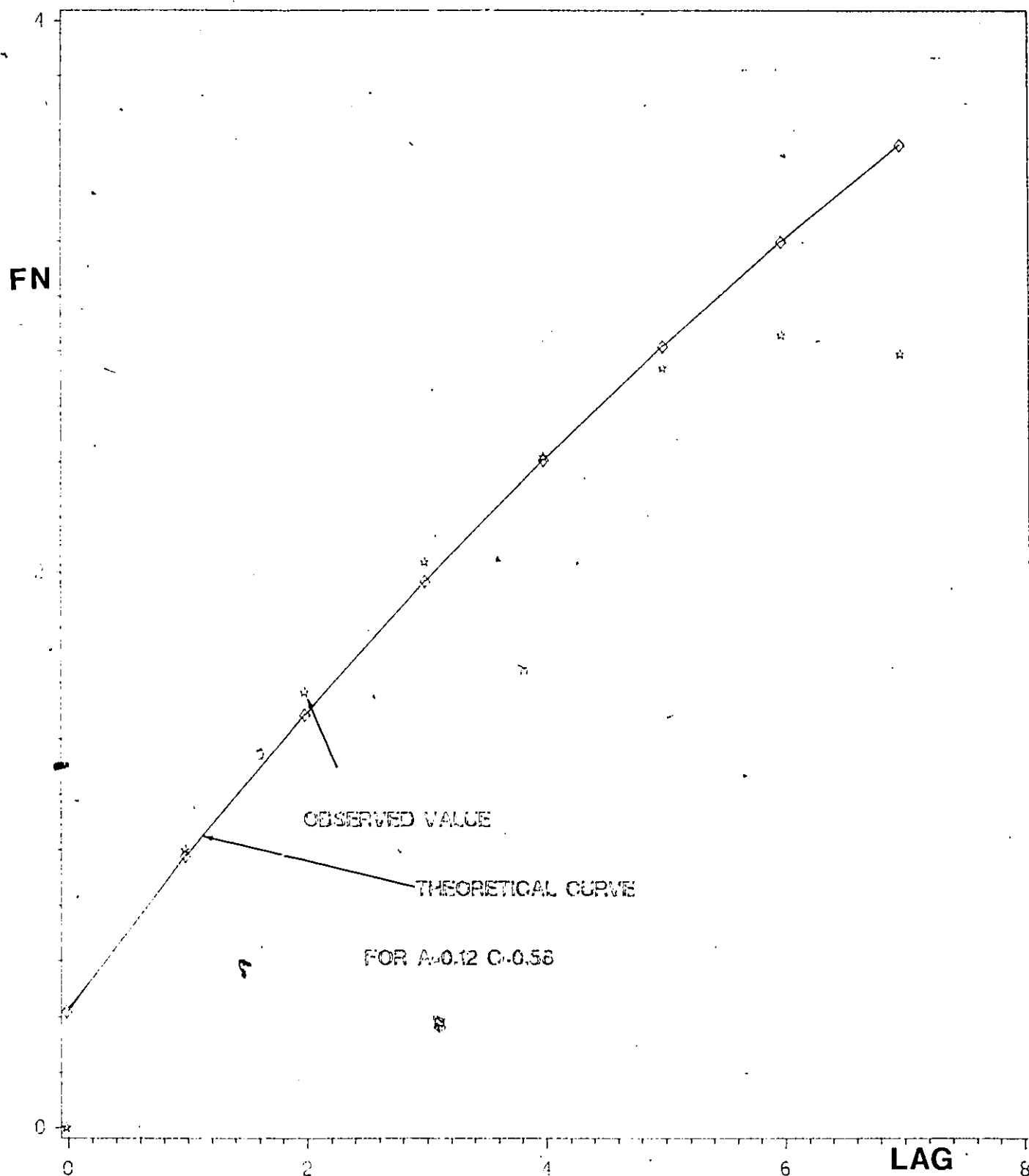


Figure 4.21e<sub>1</sub> Theoretical curve for  $a = 0.12$  and  $c = 0.58$  in comparison with observed values of B14(Agglutinants)

variances, were calculated according to Eq.59 and are shown in Tables 4.11 - 4.14. Theoretical and observed relative variances were plotted against the lag as seen in Fig. 4.21a<sub>1</sub>, 4.21b<sub>1</sub>, 4.21c<sub>1</sub>, 4.21d<sub>1</sub> and 4.21e<sub>1</sub>.

TABLE 4.11 Theoretical and observed of the relative variance of B1 (Eoguttulina sp.)

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.42
1	2.1703	1.00	0.98
2	1.6845	1.55	1.50
3	1.3982	1.93	1.97
4	1.2171	2.24	2.41
5	1.1205	2.58	2.82
6	1.0366	2.87	3.17
7	0.9719	3.13	3.51

TABLE 4.12 Theoretical and observed values of the relative variance of B2 (Epistomina mosquensis)

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.12
1	127.83	1	0.94
2	107.66	1.68	1.63

3	97.19	2.28	2.30
4	90.21	2.82	2.86
5	80.51	3.15	3.36
6	72.73	3.41	3.79
7	65.74	3.60	4.18

TABLE 4.13 Theoretical and observed values of the relative variance of B11 (Spirillina tenuissima)

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.1
1	311.50	1	0.95
2	259.93	1.66	1.72
3	246.42	2.37	2.40
4	236.71	3.04	3.04
5	221.40	3.55	3.60
6	209.80	4.04	4.12
7	193.60	4.35	4.58

TABLE 4.14 Theoretical and observed values of the relative variance of B13 (Ophthalmidium strumosum)

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.09
1	90.56	1	0.97

2	82.61	1.82	1.78
3	76.85	2.55	2.54
4	74.17	3.28	3.25
5	72.24	3.99	3.91
6	69.48	4.60	4.52
7	66.23	5.12	5.10

TABLE 4. 15 Theoretical and observed values of the relative variance of B14 (agglutinants)

$k(\text{lag})$	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.42
1	24.74	1	0.98
2	19.41	1.57	1.49
3	16.83	2.04	1.97
4	14.98	2.42	2.41
5	13.54	2.74	2.82
6	11.79	2.86	3.20
7	9.88	2.79	3.55

Taxa B1, B2, B11, B13 and B14 show a good fit between observed and theoretical relative variance curves.

As explained before, the cumulative frequency distribution tables are constructed with single, paired and sets of 4 consecutive values of each taxon respectively ( Tables 4.19, 4.21, 4.23, 4.25 and 4.27 ). Each of these cumulative distributions ( Tables 4.20, 4.22, 4.24, 4.26 and 4.28 ) was plotted on normal probability paper. R values were

estimated for single, paired and sets of four values of each taxon, using Appendix II or III, according to their means and variances as shown in Tables 4.16, 4.17 and 4.18.

For single, paired and sets of four values of each taxon Y functions were estimated using Eq. 60. These functions were plotted on normal probability paper which, as before, was used for the probit transformed

TABLE 4.16 Mean, variance, R and b parameters for species B1 ( Eoguttulina sp. ), B2 ( Epistomina mosquensis ), B11 ( Spirillina tenuissima ), B13 ( Ophthalmidium strumosum ) and B14 ( agglutinants ) for N=30 samples (single values )

Taxa	B1	B2	B11	B13	B14
$\bar{x}$	0.02847	0.1384	0.25753	0.1125	0.10416
$s^2(x)$	0.00020	0.01236	0.03011	0.00875	0.00235
R	0.23	0.50	0.51	0.51	0.24
b	1.903	1.087	0.644	1.215	1.264
N	30	30	30	30	30

TABLE 4.17 Mean, variance, R, and b parameters for species B1, B2, B11, B13 and B14 for N=15 samples (pairs values )

TAXA	B1	B2	B11	B13	B14
$\bar{x}$	0.02847	0.1384	0.25753	0.1125	0.10416
$s^2(x_2)$	0.00016	0.01038	0.02499	0.00796	0.00188

R	0.18	0.45	0.44	0.45	0.19
b	1.903	1.087	0.644	1.215	1.264
N	15	15	15	15	15

TABLE 4. 18 Mean, variance, R and b parameters for species B1, B2, B11, B13 and B14 for N=7 (sets of four adjacent values)

TAXA	B1	B2	B11	B13	B14
$\bar{x}$	0.02847	0.1384	0.25753	0.1125	0.10416
$s^2(x_4)$	0.000115	0.00871	0.02288	0.00718	0.00145
R	0.16	0.43	0.41	0.44	0.17
b	1.903	1.087	0.644	1.215	1.264
N	7	7	7	7	7

Experimental frequency distributions for five samples of single, paired and sets of 4 microfossil values were plotted on two normal probability scales. The values of b and R were obtained from the means and variances of the microfossil taxon proportion values using Eq. 60 and 62. They determine the positions and slopes of the prob - prob lines which satisfy Eq. 60 as shown in Figs. 4.22 a, b, c - 4.23 a, b, c - 4.24 a, b, c - 4.25 a, b, c and 4.26 a, b, c.

Table 4.19 Probit transformation of B1 (Eoguttulina sp.) from Tojeira 2 section.

sample single %	probit single	sample pairs %	probit pairs	sample set of 4 %	probit set of 4
4.4	3.294				
2.4	3.023	3.4	3.175		
3.8	3.226				
2.3	3.005	3.05	3.120	3.225	3.165
4.2	3.272				
4.8	3.335	4.5	3.305		
3.8	3.226				
3.1	3.134	3.45	3.181	3.975	3.241
4.2	3.272				
3.0	3.119	3.6	3.315		
4.1	3.261				
2.8	3.089	3.45	3.181	3.525	3.191
2.7	3.073				
4.2	3.272	3.45	3.181		
6.9	3.517				
3.4	3.175	5.15	3.369	4.300	3.283
2.5	3.040				
1.1	2.710	1.8	2.903		
3.6	3.201				

( continued )

Table 4.19 ( concluded )

3.7	3.213	3.65	3.415	2.725	3.077
2.2	2.986				
1.9	2.925	2.05	2.956		
1.7	2.880				
0.4	2.348	1.05	2.692	1.55	3.077
2.4	3.023				
0.7	2.543	1.55	2.843		
0.4	2.348				
1.4	2.803	0.90	2.634	1.225	2.129
1.4	2.803				
1.9	2.925	1.65	2.868		

A

Table 4.20 Cumulative frequency distributions of B1 (Eoguttulina sp.) for 30, 15 and 7 samples.

N = 30

class int.	freq.	cum. freq.	perc.	cum. per.	plotting value
2.375 - 2.625	3	3	10.0	10.0	8.80
2.625 - 2.875	3	6	10.0	20.0	18.70
2.875 - 3.125	11	17	36.67	56.67	54.94
3.125 - 3.375	12	29	40.0	96.67	94.50
3.375 - 3.625	1	30	3.33	100.0	97.80

N = 15

2.70 - 2.90	4	4	26.67	26.67	23.91
2.90 - 3.10	2	6	13.33	40.00	36.96
3.10 - 3.30	5	11	33.33	73.33	69.56
3.30 - 3.50	4	15	26.67	100.0	95.65

N = 7

2.20 - 2.60	1	1	14.29	14.29	9.10
2.60 - 3.00	1	2	14.29	28.57	22.73
3.00 - 3.40	5	7	71.43	100.0	90.91

Table 4.21 Probit transformation of B2 ( Epistomina mosquensis ) from Tojeira 2 section.

sample single %	probit single	sample pairs %	probit pairs	sample set of 4 %	probit set of 4
0.0	2.187				
1.9	2.925	0.95	2.654		
1.0	2.674				
0.0	2.187	0.5	2.424	0.50	2.424
1.1	2.710				
1.6	2.856	1.35	2.788		
2.4	3.023				
5.2	3.374	3.80	3.226	2.575	3.053
0.5	2.424				
0.0	1.187	0.25	2.137		
1.8	2.903				
19.0	4.122	10.4	3.741	5.325	3.386
17.9	4.081				
26.3	4.366	22.1	4.231		
16.6	4.030				
13.8	3.911	15.2	3.972	18.65	4.109
19.0	4.122				
18.6	4.107	18.8	4.115		
16.4	4.100				

( continued )

Table 4.21 ( concluded )

6.7	3.501	11.55	3.802	15.175	3.971
31.1	4.507				
16.7	4.034	23.9	4.290		
10.9	3.768				
23.2	4.268	17.05	4.048	20.475	4.175
22.8	4.255				
21.5	4.211	22.15	4.233		
37.6	4.687				
24.9	4.322	31.25	4.511	26.7	4.378
31.7	4.524				
25.0	4.326	28.35	4.427		

V

Table 4.22 Cumulative frequency distributions of B2 (Epistomina mosquensis) for 30, 15 and 7 samples.

N = 30

class interval	freq	cum. freq.	perc. %	cum. %	plotting value
2.25 - 2.75	6	6	20.00	20.00	18.70
2.75 - 3.25	4	10	13.33	33.33	32.00
3.25 - 3.75	2	12	6.67	40.00	38.50
3.75 - 4.25	10	22	33.33	73.33	71.43
4.25 - 4.75	8	30	26.67	100.0	97.80

N = 15

2.10 - 2.70	3	3	20.00	20.00	17.40
2.70 - 3.30	2	5	13.33	33.33	30.40
3.30 - 3.90	2	7	13.33	46.67	43.50
3.90 - 4.50	8	15	53.33	100.0	95.65

N = 7

2.70 - 3.30	2	2	28.57	28.57	22.80
3.30 - 3.90	1	3	14.29	42.86	36.40
3.90 - 4.50	4	7	57.14	100.0	90.90

Table 4.23 Probit transformation of B11 ( *Spirillina tenuissima* )  
for N= 30, 15 and 7 samples from the Tojeira 2 section.

sample single %	probit single	sample pairs %	probit pairs	sample set of 4 %	probit set of 4
57.7	5.194				
24.0	4.294	40.85	4.768		
44.8	4.869				
61.0	5.279	52.9	5.073	46.875	4.921
42.0	4.798				
64.1	5.361	53.05	5.076		
38.6	4.710				
30.9	4.501	34.75	4.608	43.90	4.846
46.8	4.920				
52.7	5.068	49.75	4.993		
46.8	4.920				
19.4	4.137	33.1	4.563	41.425	4.784
14.9	3.969				
15.0	3.964	14.95	3.961		
13.8	3.911				
16.2	4.014	15.00	3.964	14.975	3.963
17.0	4.046				
16.7	4.034	16.85	4.040		
12.5	3.850				

( continued )

Table 4.23 ( concluded )

11.9	3.820	12.20	3.835	14.525	3.943
17.6	4.069				
11.4	3.794	14.5	3.942		
15.2	3.972				
16.4	4.022	15.8	3.997	15.15	3.970
13.4	3.892				
7.3	3.546	10.35	3.738		
10.4	3.741				
8.8	3.647	9.60	3.695	9.975	3.717
11.8	3.815				
13.5	3.897	12.65	3.856		

Table 4.24 Cumulative frequency distributions of B11 (Spirillina tenuissima) for 30, 15 and 7 samples.

N = 30

class interval	freq.	cum. freq.	perc. %	cum. %	plotting value
3.40 - 3.80	4	4	13.33	13.33	12.10
3.80 - 4.20	15	19	50.00	63.33	61.50
4.20 - 4.60	2	21	6.67	70.00	68.10
4.60 - 5.00	5	26	16.67	86.67	84.60
5.00 - 5.40	4	30	13.33	100.0	97.80

N = 15

3.75 - 4.05	9	9	60.0	60.0	56.50
4.05 - 4.35	0	9	00.0	60.0	56.50
4.35 - 4.65	2	11	13.33	73.33	69.56
4.65 - 4.95	4	15	26.67	100.0	95.65

N = 7

3.80 - 4.20	4	4	57.14	57.14	50.00
4.20 - 4.60	0	4	0.00	57.14	50.00
4.60 - 5.0	3	7	42.86	100.0	90.91

Table 4.25 Probit transformation of B13 ( Ophthalmidium strumosum )  
for 30, 15 and 7 samples from Tojeira 2 section.

sample single %	probit single	sample pairs %	probit pairs	sample set of 4 %	probit set of 4
0.2	2.122				
0.5	2.424	0.35	2.300		
0.0	2.187				
0.0	2.187	0.00	2.187	0.175	2.069
1.1	2.710				
0.0	2.187	0.55	2.456		
1.0	2.674				
0.9	2.634	0.95	2.654	0.75	2.567
2.8	3.089				
0.4	2.348	1.60	2.856		
1.8	2.903				
5.2	3.374	3.50	3.188	2.55	2.456
14.9	3.959				
16.3	4.018	15.6	3.989		
22.4	4.241				
16.7	4.034	19.55	4.156	17.575	4.068
12.0	3.825				
19.4	4.134	15.7	3.993		
27.5	4.402				

( continued )

Table 4.25 ( concluded )

14.8	3.955	21.15	4.198	18.425	4.101
16.5	4.026				
27.4	4.399	21.95	4.226		
29.8	4.470				
16.1	4.010	22.95	4.259	22.45	4.242
18.3	4.096				
20.9	4.190	19.60	4.144		
11.5	3.800				
8.3	3.615	9.90	3.713	14.75	3.954
15.4	3.981				
15.4	3.981	15.4	3.981		

Table 4.26 Cumulative frequency distributions of B13 ( Ophthalmidium strumosum ) for 30, 15 and 7 samples.

N = 30

class interval	freq.	cum. freq.	perc. %	cum. %	plotting value
2.25 - 2.75	9	9	30.00	30.00	28.60
2.75 - 3.25	2	11	6.67	36.67	35.20
3.25 - 3.75	2	13	6.67	43.33	41.80
3.75 - 4.25	14	27	46.67	90.00	88.00
4.25 - 4.75	3	30	10.00	100.0	97.80

N = 15

2.25 - 2.75	4	4	26.67	26.67	23.90
2.75 - 3.25	2	6	13.33	40.0	37.00
3.25 - 3.75	1	7	6.67	46.67	41.80
3.75 - 4.25	8	15	53.33	100.0	95.70

N = 7

2.0 - 2.8	3	3	42.86	42.86	36.40
2.8 - 3.6	0	3	0.00	42.86	36.40
3.6 - 4.4	4	7	57.14	100.0	90.90

Table 4.27 Probit transformation of B14 ( agglutinants ) for N = 30, 15 and 7 samples from Tojeira 2 section.

sample single	probit single	sample pairs	probit pairs	sample set of 4	probit set of 4
7.5	3.560				
3.9	3.238	5.7	3.420		
3.5	3.188				
3.3	3.162	3.4	3.175	4.55	3.331
4.2	3.283				
2.4	3.023	3.3	3.162		
12.2	3.835				
7.3	3.546	9.75	3.704	6.525	3.488
7.9	3.588				
7.1	3.532	7.5	3.560		
6.6	3.494				
11.2	3.784	8.9	3.653	8.20	3.608
14.9	3.959				
13.4	3.892	14.15	3.926		
5.6	3.411				
14.8	3.955	10.2	3.730	12.175	3.834
13.5	3.897				
10.3	3.735	11.9	3.820		
9.3	3.667				
7.4	3.553	8.35	3.618	10.125	3.725
11.0	3.773				

( continued )

Table 4.27 ( concluded )

16.7	4.034	13.85	3.913		
11.8	3.815				
10.5	3.746	11.15	3.781	12.5	3.850
17.8	4.077				
20.8	4.187	19.3	4.133		
12.3	3.840				
20.9	4.190	16.6	4.030	17.95	4.083
13.7	3.906				
10.7	3.757	12.2	3.835		

Table 4.28 Cumulative frequency distributions of agglutinants for 30, 15 and 7 samples.

N=30

class interval	freq.	cum. freq	perc. %	cum. %	plotting value
3.10 - 3.30	5	5	16.67	16.67	15.40
3.30 - 3.50	2	7	6.67	23.33	22.00
3.50 - 3.70	6	13	20.0	43.33	41.80
3.70 - 3.90	10	23	33.33	76.67	74.72
3.90 - 4.10	7	30	23.33	100.0	97.80

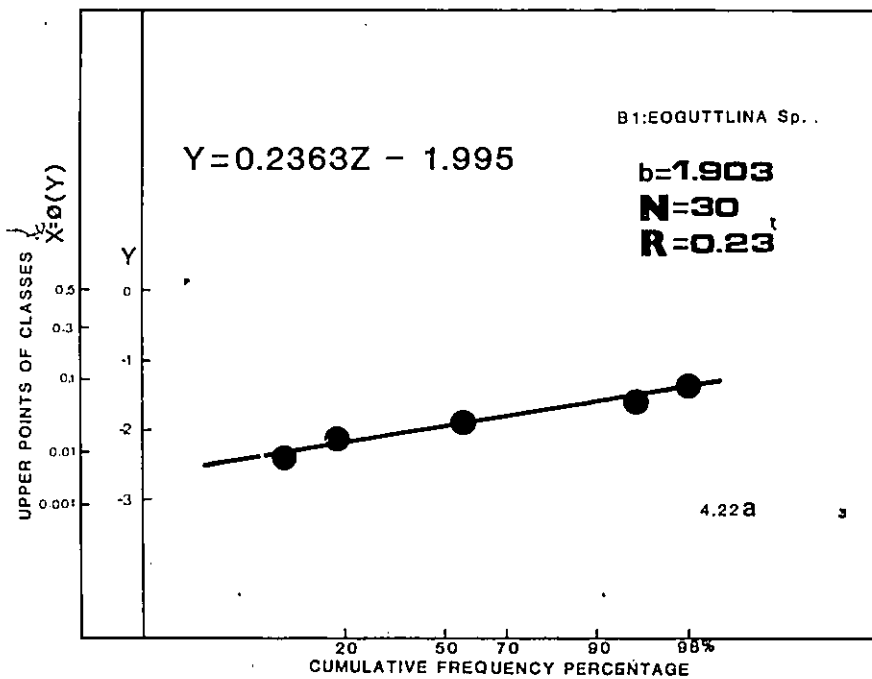
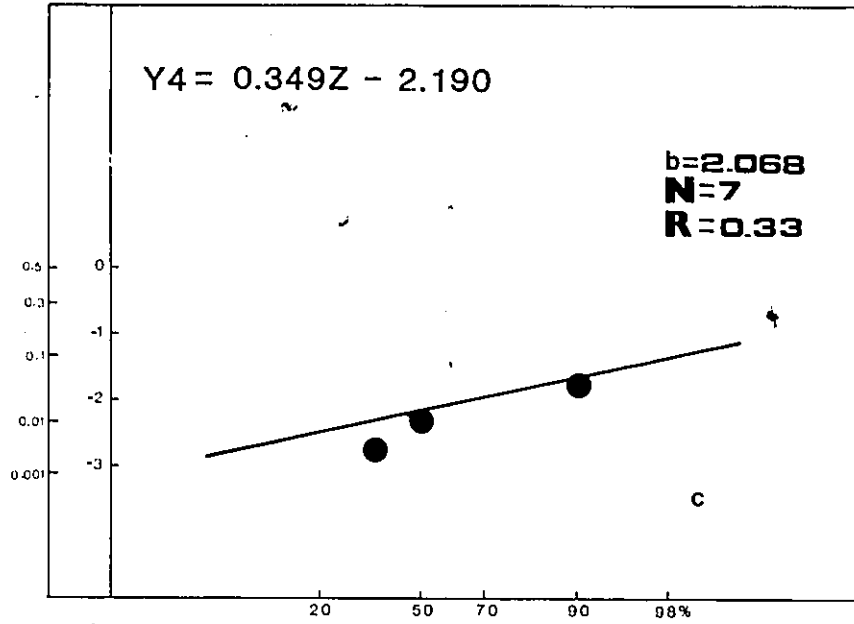
N = 15

3.15 - 3.45	3	3	20.00	20.00	17.40
3.45 - 3.75	5	8	33.33	53.33	50.00
3.75 - 4.05	5	13	33.33	86.67	82.60
4.05 - 4.35	2	15	13.33	100.0	95.65

N = 7

3.375 - 3.625	3	3	42.86	42.86	36.40
3.625 - 3.875	3	6	42.86	85.71	77.30
3.875 - 4.125	1	7	14.29	100.0	90.90

Figure 4.22a,b,c Experimental frequency distributions for 30,15 and 7 values of Eoguttulina sp. and their theoretical frequency distributions which are probnormal distributions



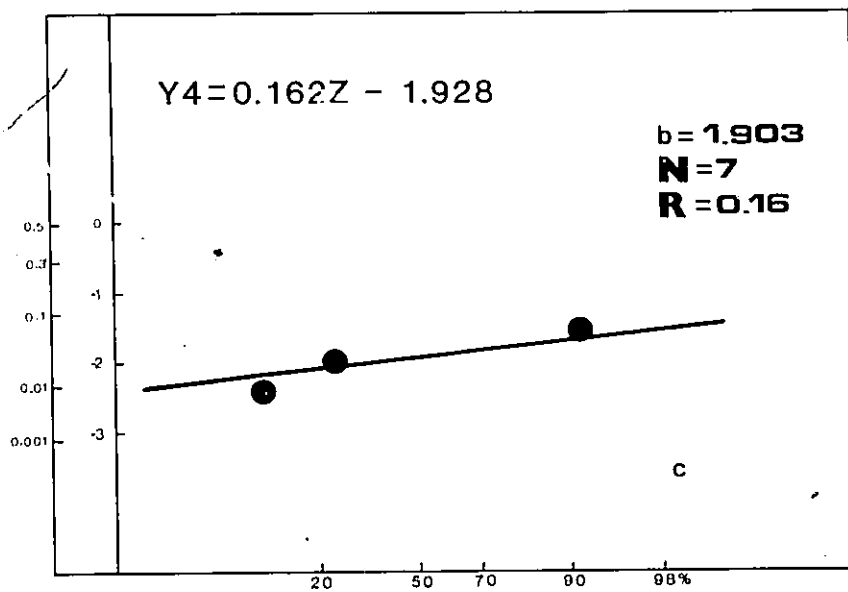
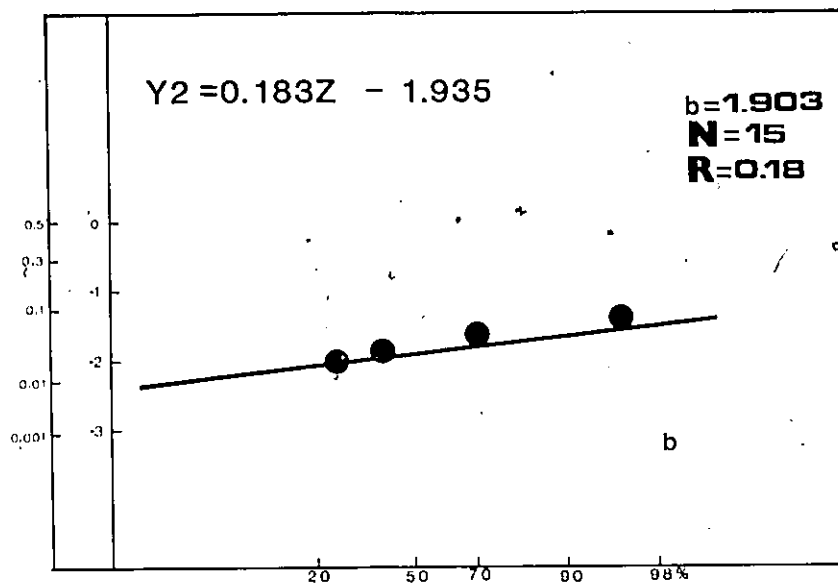


Figure 4.23a,b,c Experimental frequency distributions for 30,15 and 7 values of E. mosquensis and their theoretical frequency distributions satisfying Eq.60

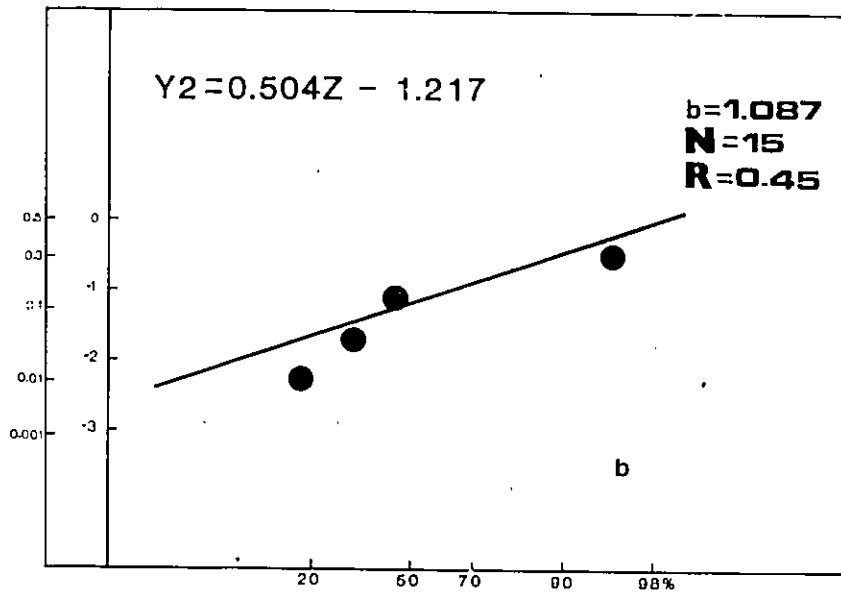
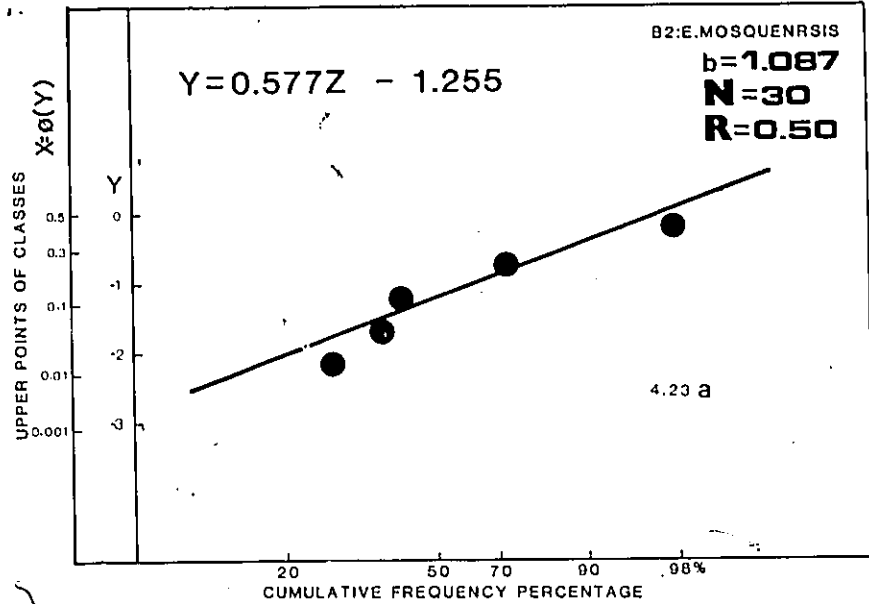
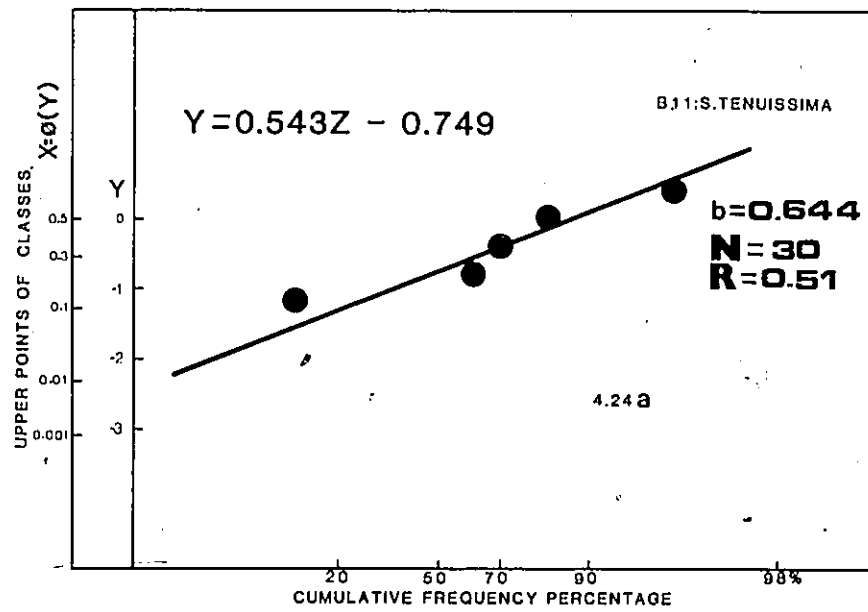
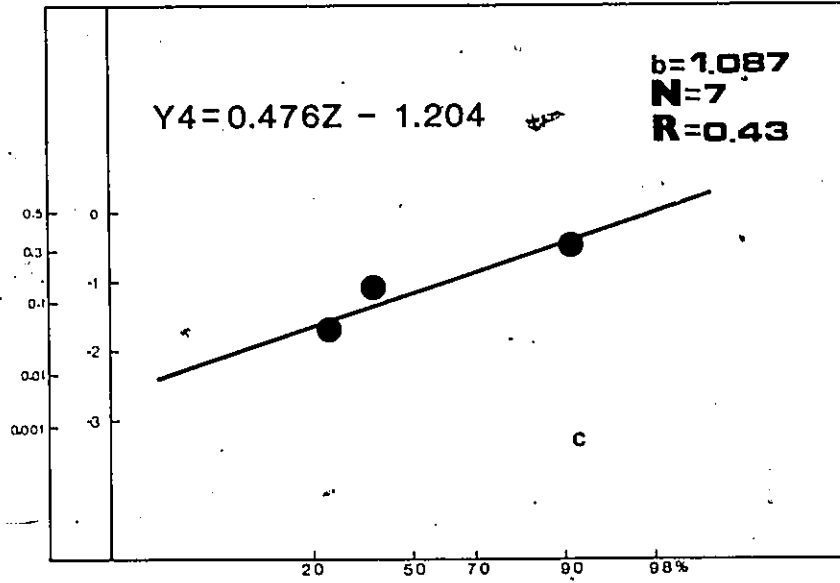


Figure 4.24a,b,c Experimental and theoretical distributions  
for 30,15 and 7 values of S. tenuissima



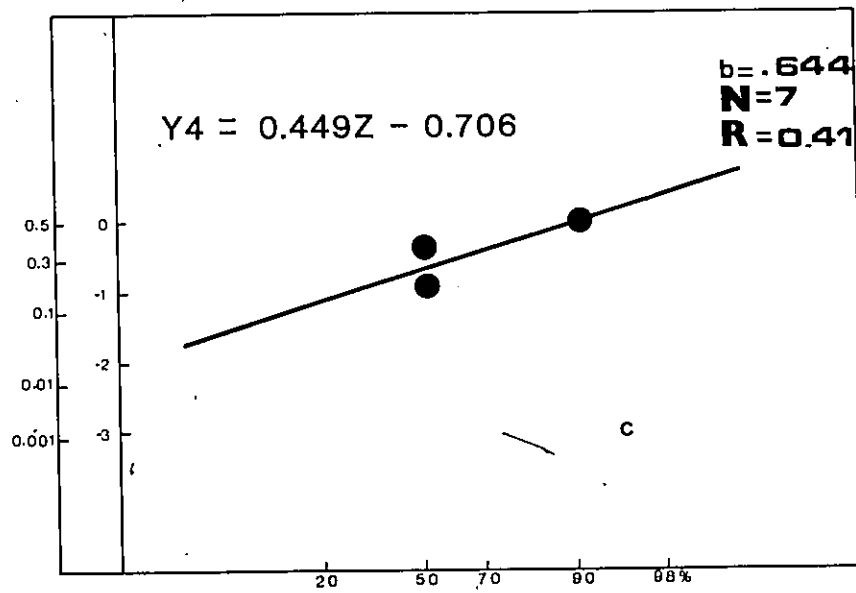
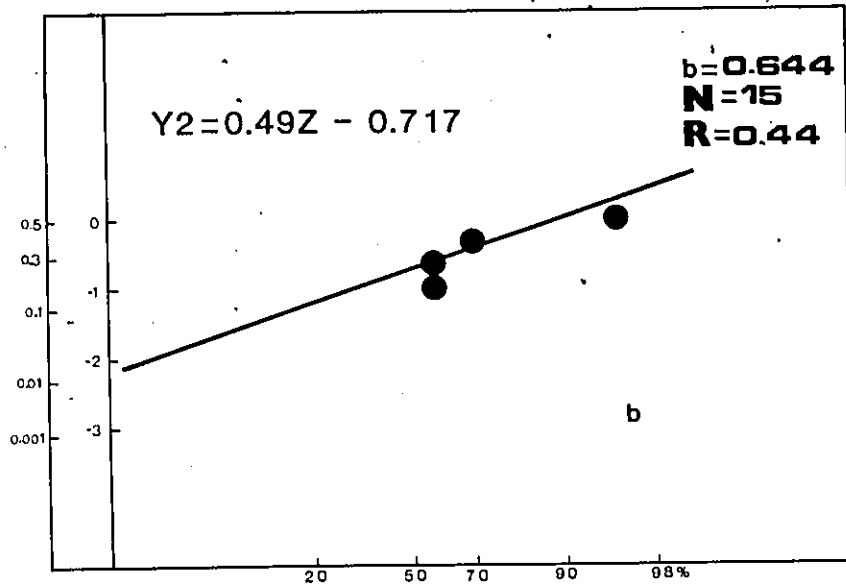


Figure 4.25a,b,c Experimental and theoretical distributions  
for 30,15 and 7 values of O. strumosum

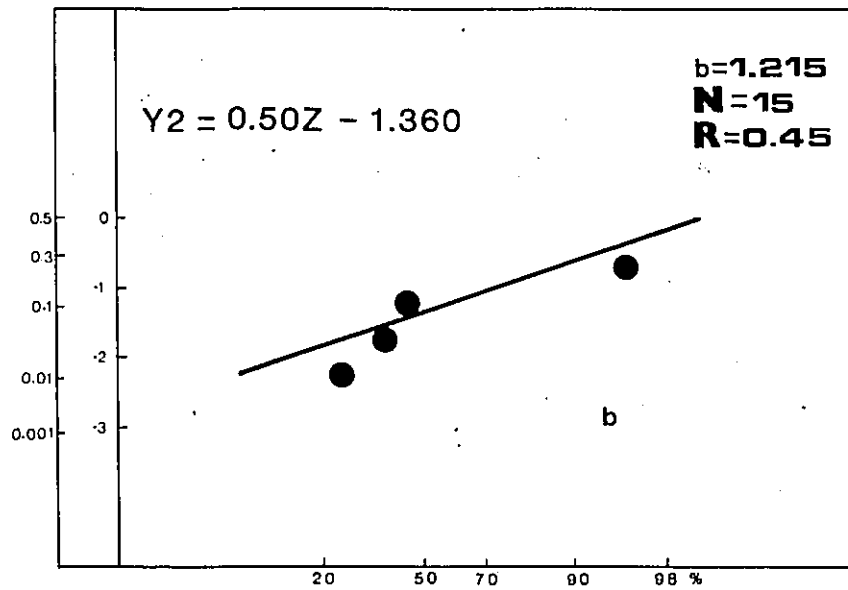
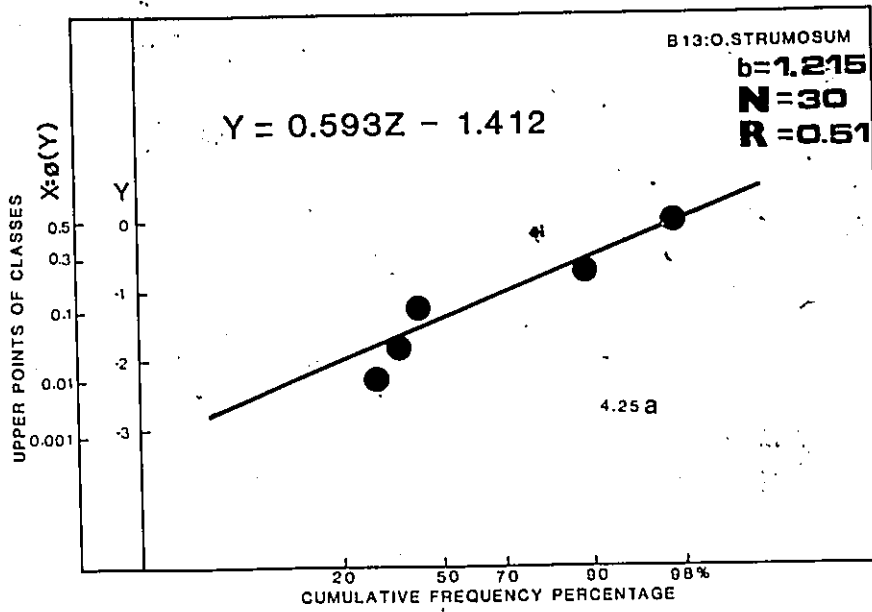
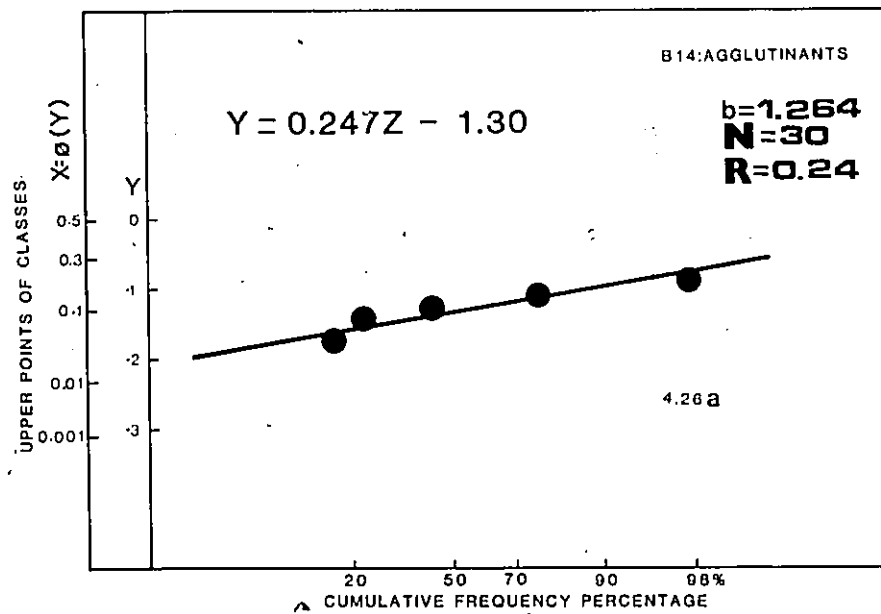
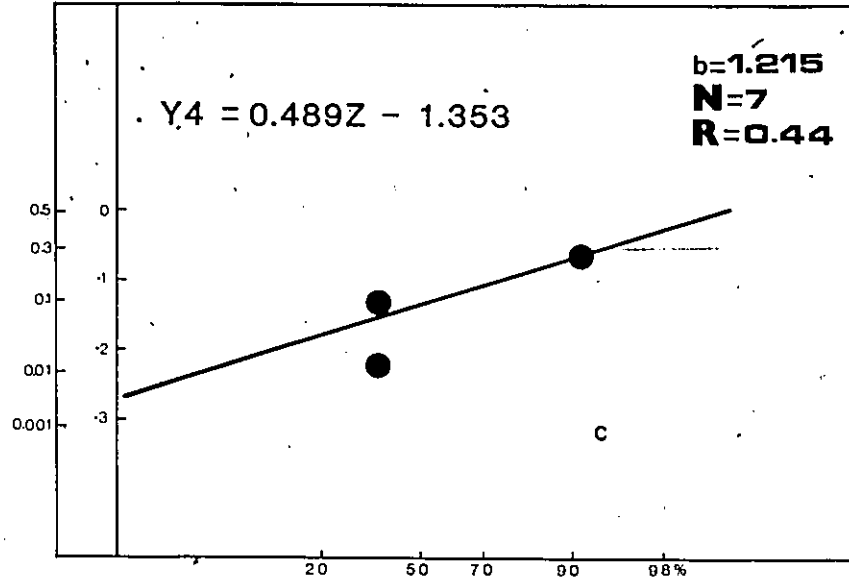
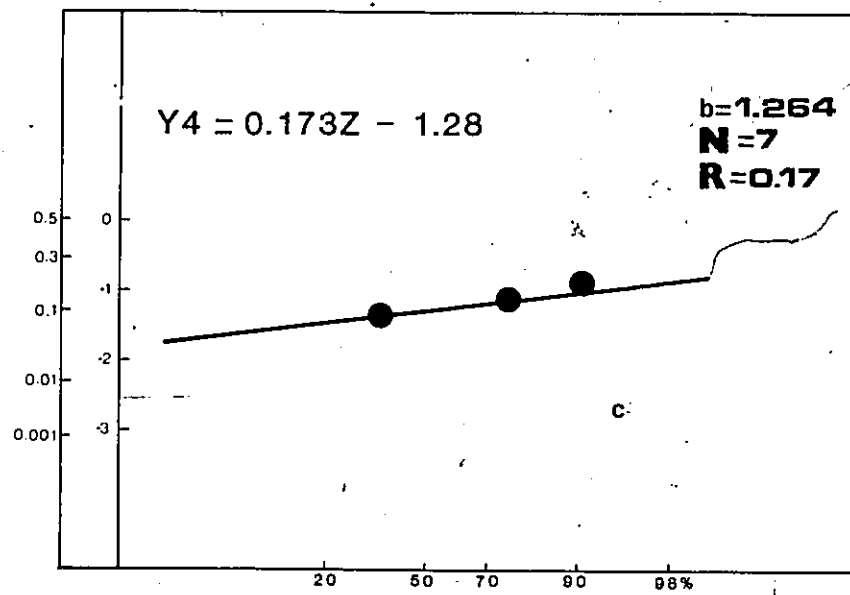
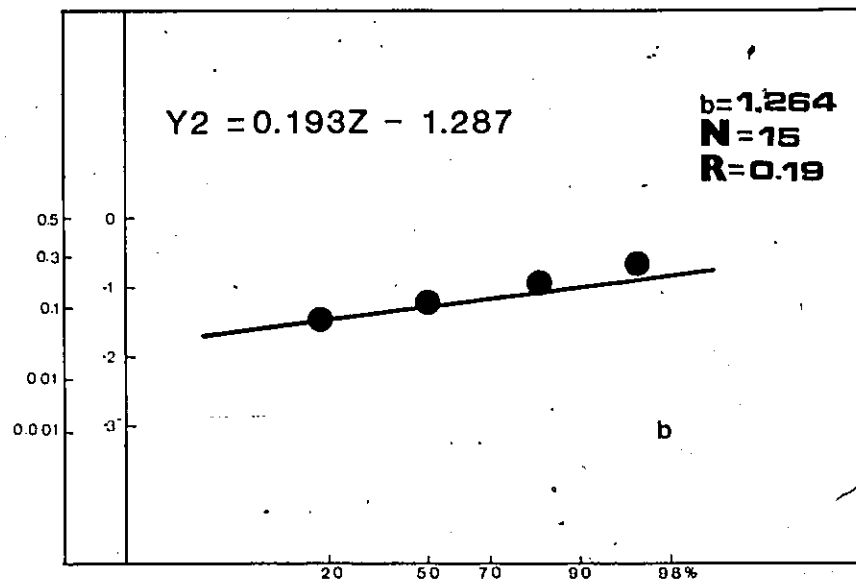


Figure 4.26a,b,c Experimental and theoretical distributions  
for 30,15 and 7 values of Agglutinants





## CHAPTER 5 New data set and geostatistical modelling

### 5.1 Introduction

The Tojeira 1 and Tojeira 2 sections were resampled by this author and F. M. Gradstein during the summer of 1986.

### 5.2 - Tojeira 1 section

The samples were collected at regular intervals of 5 m. The sampled section starts just below outcrop 17 of Stam ( 1986). A total of 23 samples were collected, with 5 of these samples coming from the same layer.

As explained before, with only 17 observations, there were less than the minimum ( 30 ) recommended for initiating an ARIMA procedure. About 4 autocorrelations would be a safe number to examine, since this is one - fourth of the number of observations. However, this is still not sufficient to draw decisive conclusions about the microfossil abundance.

When ARIMA procedures were applied the autocorrelations of the species turned out to be not significantly different from zero at the 5% significance level. The null hypothesis of absence of autocorrelation was acceptable. All time series of species were white noise. This was expected because 17 samples are fewer than the necessary number for the ARIMA procedure to be considered accurate.

### 5.3 Tojeira 2 section

A total of 13 samples were recollected from the Tojeira 2 section. They were not collected at a regular interval as explained before. An ARIMA procedure could not be applied to this data set either, because of the small sample size.

### 5.4 - Correlation of Stam's and Gradstein's data

The new data set was not sufficiently large to construct time series models for either Tojeira 1 or Tojeira 2 separately. To overcome this problem a larger data set was created by combining Stam's and Gradstein's data using samples taken at different locations only. Before creating the enlarged data set, Stam's and Gradstein's data, which were collected at approximately the same location, were correlated to each other. The idea was to see how close the data from the two different data sets related to each other ( Tables 5.1 and 5.2).

#### 5.4 1 - Tojeira 1 section:

Table 5.1 was used as input for the SAS CORR procedure to calculate correlation coefficients.

As can be seen in Table 5.3, Ophthalmidium strumosum and the taxa Spirillina elongata and Eoguttulina have a high positive correlation. The taxa Epistomina uhligi, lenticulina, Epistomina mosquensis,



Table 5.2 Comparison of Stam's and Gradstein's data for Tojeira 2 section

Sample No.	B1		B2		B3		B4		B5		B6		B7	
	G	S	G	S	G	S	G	S	G	S	G	S	G	S
T2/1	0.4	2.4	0.0	1.9	6.0	3.4	17.8	5.2	4.8	1.9	0.6	5.3	0.4	3.9
T2/2	0.6	4.3	0.0	0.6	3.3	0.0	0.8	0.0	8.0	2.3	1.1	2.7	0.8	1.8
T2/3	4.5	3.8	0.0	4.0	0.7	3.0	7.0	0.0	4.5	4.6	2.7	8.2	1.0	2.3
T2/5	0.5	3.2	1.0	2.0	2.8	0.8	0.9	0.0	4.8	5.2	7.7	5.2	0.4	4.0
T2/6	0.5	3.2	0.0	0.6	3.3	0.3	6.0	0.0	3.9	4.9	7.1	2.8	0.9	2.8
T2/8	0.0	3.8	1.5	1.6	3.3	3.5	8.0	0.0	2.3	1.3	3.3	4.2	0.2	3.0
T2/9	0.0	3.2	2.0	1.3	1.0	3.5	4.0	0.0	3.2	5.2	2.0	8.0	0.2	4.0
T2/10	0.0	3.6	3.5	1.1	3.1	1.1	0.0	0.0	3.3	2.1	2.7	8.0	0.5	2.8
T2/11	0.3	2.1	3.1	1.0	3.1	1.1	7.5	0.0	1.3	3.5	2.7	2.2	0.0	0.3
T2/12	0.0	1.4	2.5	1.1	3.1	1.1	8.0	0.0	1.3	5.2	2.0	2.2	0.0	0.3
T2/13	0.0	1.4	3.5	1.3	1.6	1.1	5.0	0.0	1.3	2.2	5.5	2.2	0.0	0.3
	B8		B9		B10		B11		B12		B13		B14	
	G	S	G	S	G	S	G	S	G	S	G	S	G	S
	0.0	0.0	1.8	2.2	1.3	0.9	2.4	0.0	1.3	4.2	0.0	1.3	4.3	3.3
	0.0	0.0	7.0	4.1	3.1	3.8	6.3	0.0	6.5	5.7	0.0	7.0	3.2	2.2
	0.0	0.0	0.0	3.2	1.0	2.4	3.8	0.0	3.2	8.0	0.0	2.0	2.0	2.2
	0.0	0.0	0.0	4.0	3.8	4.8	5.5	0.0	3.2	7.8	0.0	2.0	7.7	2.6
	0.0	0.0	0.0	1.8	0.8	1.3	1.5	0.0	3.1	0.7	0.0	1.5	1.5	1.8
	0.0	0.0	0.0	5.4	3.0	3.0	1.5	0.0	4.4	0.7	0.0	2.5	3.0	5.8
	0.0	0.0	0.0	2.2	1.0	1.0	1.5	0.0	1.4	2.3	0.0	1.9	1.1	1.7
	0.0	0.0	0.0	3.2	1.0	1.0	1.5	0.0	1.4	0.9	0.0	1.9	1.1	1.7
	0.0	0.0	0.0	5.6	3.3	3.3	1.5	0.0	4.6	0.7	0.0	2.5	3.0	5.9
	0.0	0.0	0.0	2.2	1.0	1.0	1.5	0.0	1.4	2.3	0.0	1.9	1.1	1.7
	0.0	0.0	0.0	3.2	1.0	1.0	1.5	0.0	1.4	0.9	0.0	1.9	1.1	1.7





Lenticulina muensteri, Pseudolamarckina rjansanensis have a low positive correlation. Spirillina tenuissima, Ophthalmidium carinatum and agglutinants have positive or negative correlation coefficients that are close to zero, indicating no correlation between Stam's and Gradstein's data for these taxa. Nodosaria/Dentalina are negatively correlated. The taxa show an inverse relationship in Tojeira 1 section. There is no correlation between Stam's and Gradstein's data for Epistomina sp.. These relationships are shown graphically in Fig. 5.1.

#### 5.4.2 Tojeira 2 section

Epistomina mosquensis, Epistomina uhligi and Spirillina tenuissima have a high positive correlation. Spirillina infima, Spirillina elongata, Epistomina sp., Lenticulina sp., Ophthalmidium strumosum and Pseudolamarckina rjansanensis have a low positive correlation. There is no correlation between Stam's and Gradstein's data for Lenticulina muensteri, agglutinants and Nodosaria/Dentalina as shown in Table 5.4. These relationships are shown graphically in Fig. 5.2.

In Tables 5.3 and 5.4, the first number is a correlation coefficient and the second number is probability. The significance probability of the correlation or  $PROB > D$  is the significance probability of Hoeffding's  $D$ . This probability is approximate for Spearman, Kendall and Hoeffding statistics.

Stam's and Gradstein's data are only weakly correlated. This is probably due to the relatively large noise component in the data

which is caused by counting errors as well as local random variability. However, Eoguttulina sp., Epistomina mosquensis and Ophthalmidium strumosum showed correlations varying from low to high in the Tojeira 1 section. The taxa Epistomina uhligi, Epistomina mosquensis and Spirillina tenuissima showed a positive correlation in Tojeira 2. So, these species could be expected to reinforce the autocorrelation structure in the enlarged data set.

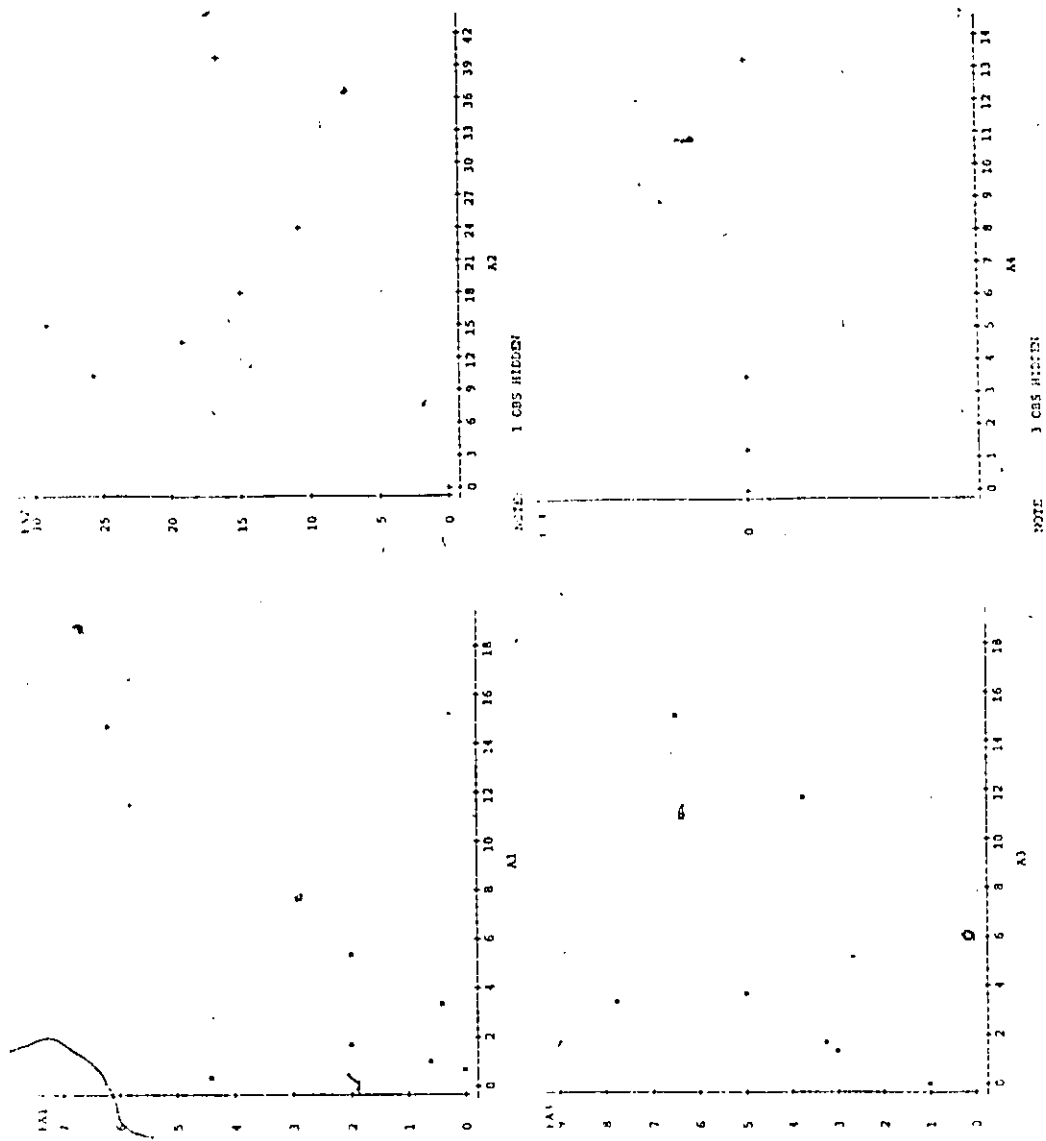


Figure 5.1 Correlation patterns of Stam's and Gradstein's data for Tojeira 1 section.

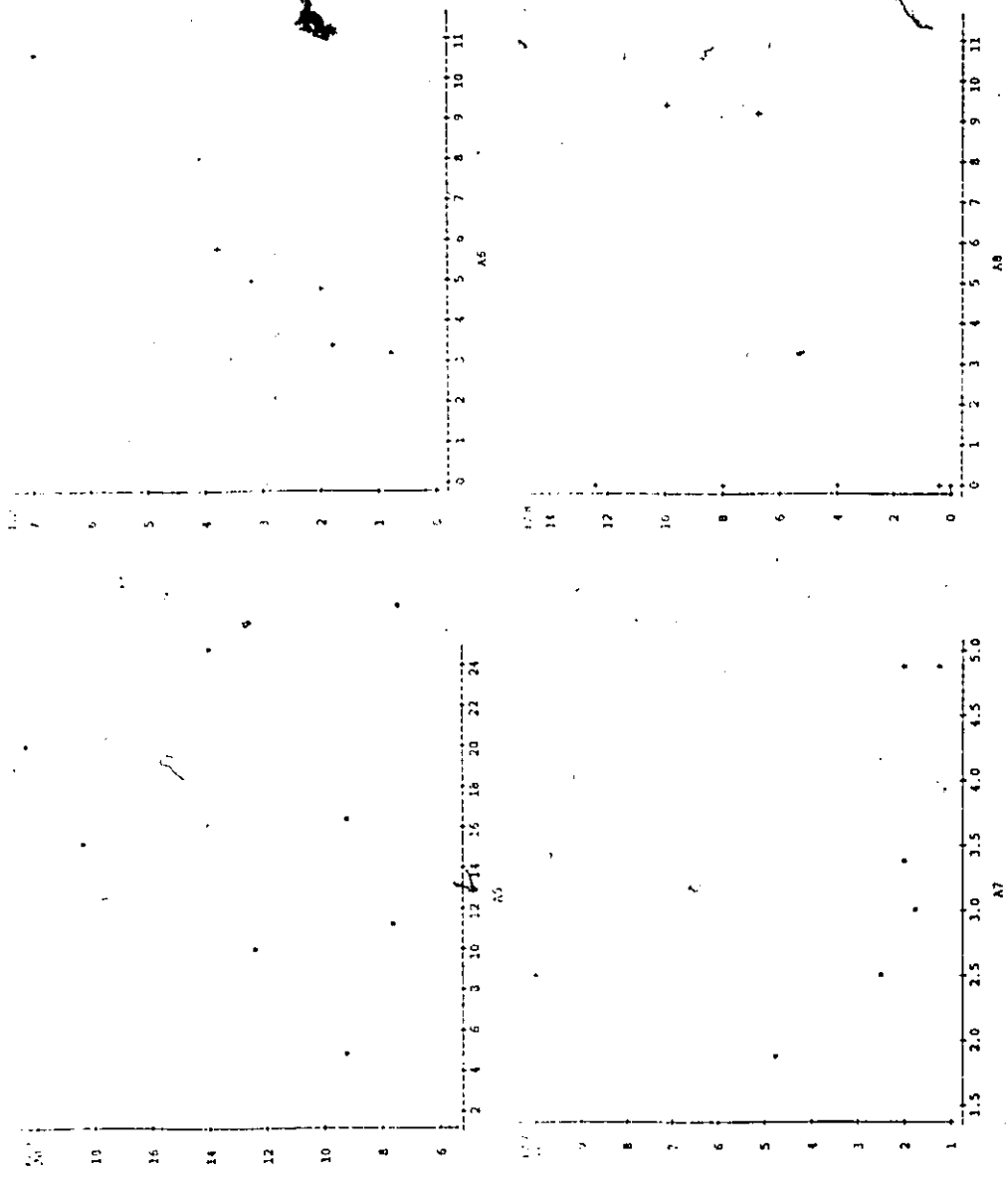


Fig. 5.1 (cont.)

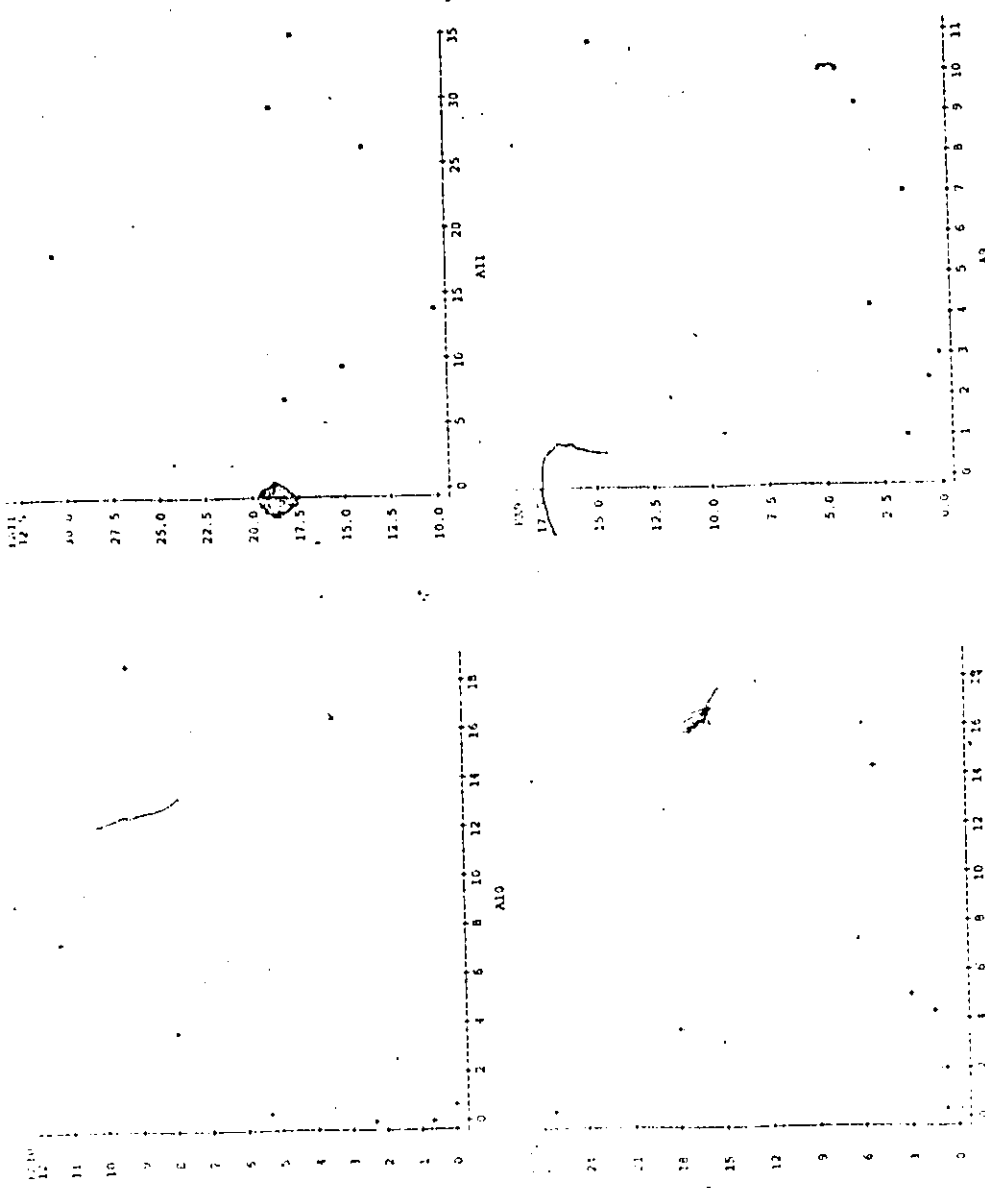


Fig. 5.1 (cont.)

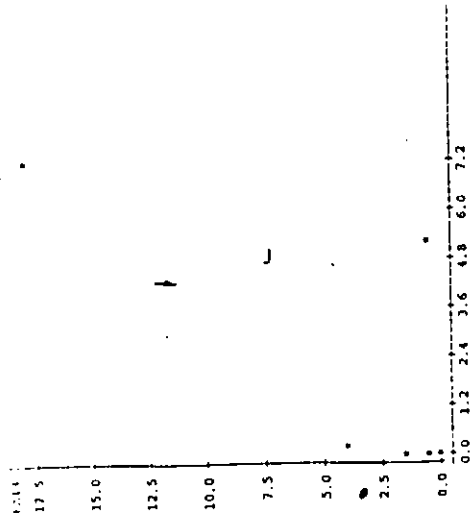
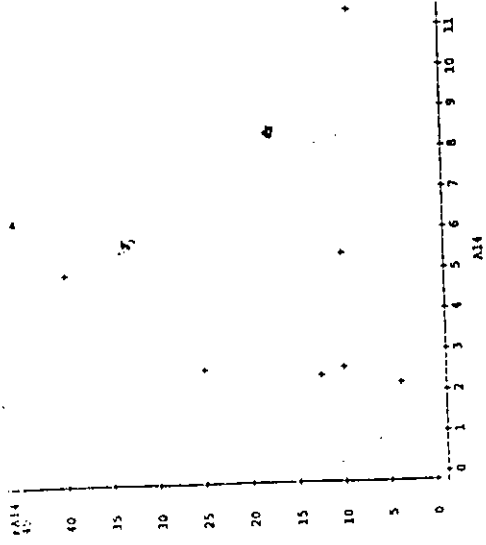


Fig. 5.1 (concluded)

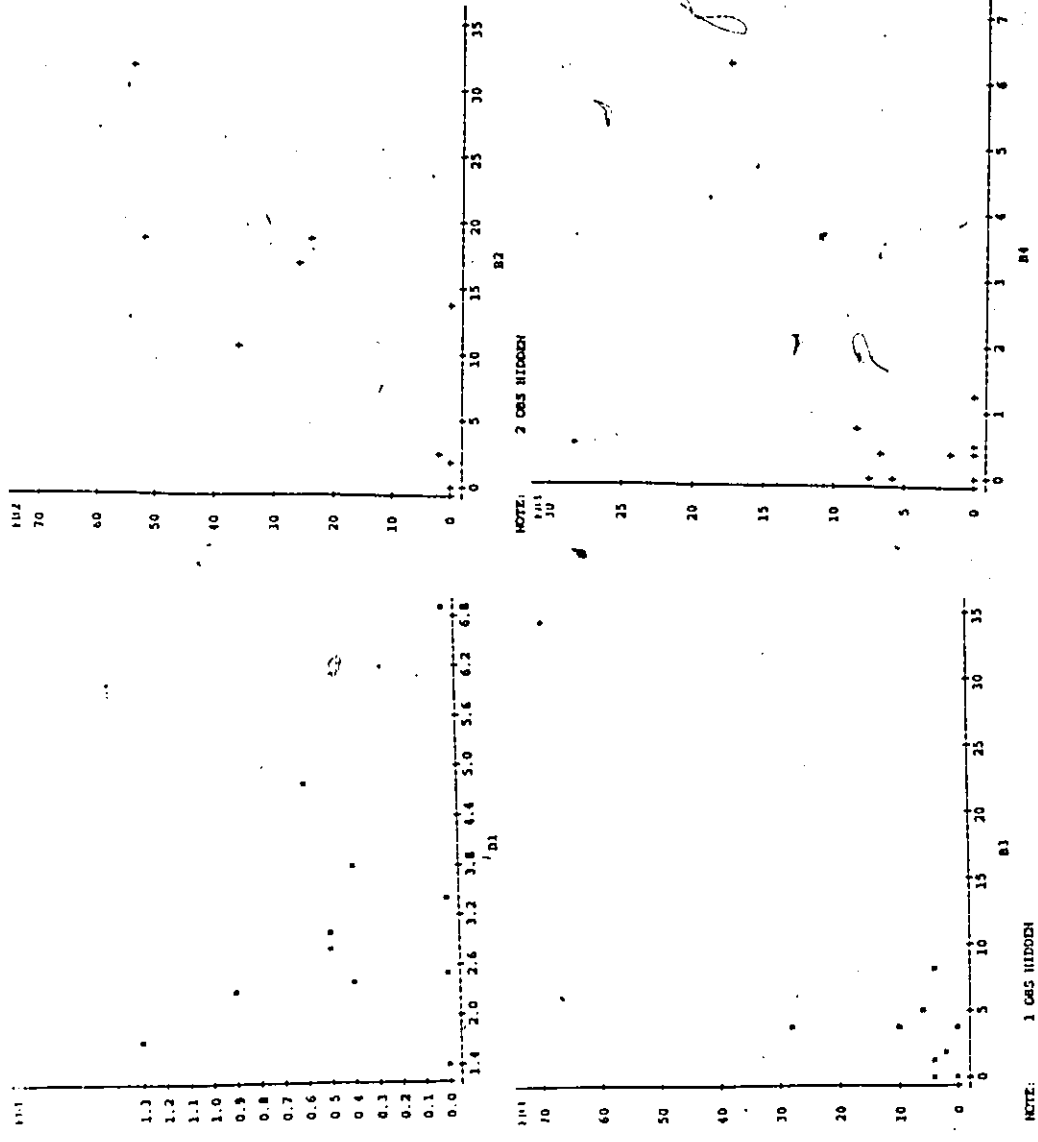
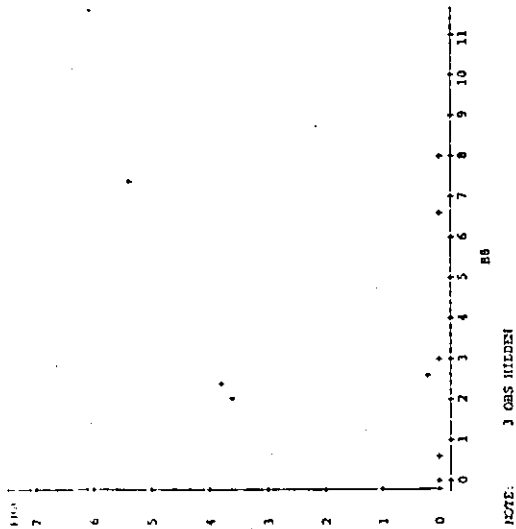
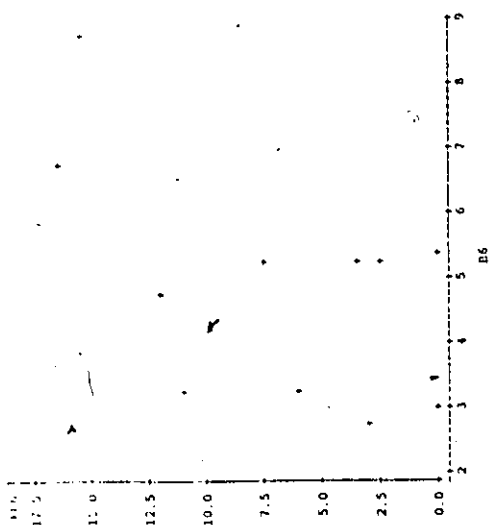


Figure 5.2 Correlation patterns of Stam's and Gradstein's data for Tojeira 2 section



NOTE: 3 OBS HIDDEN

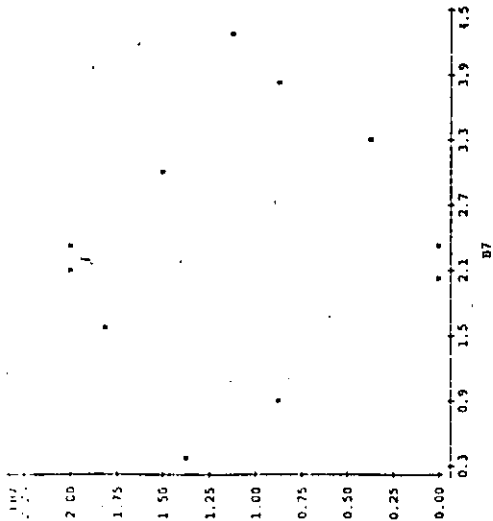
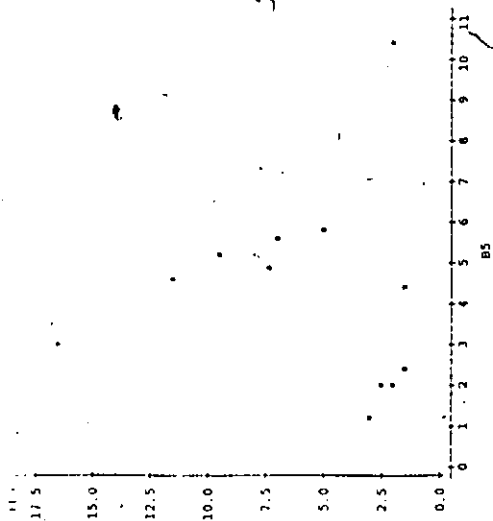


Fig. 5.2 (cont.)

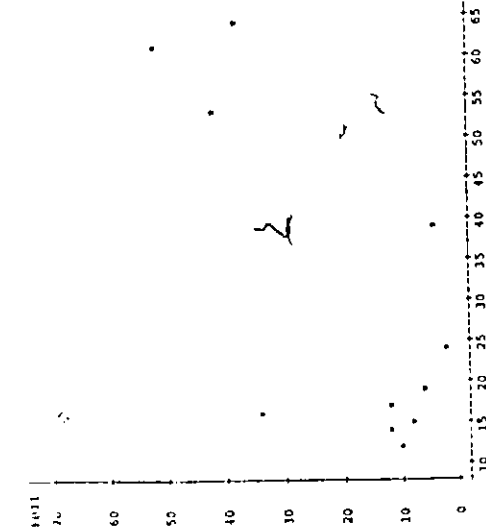
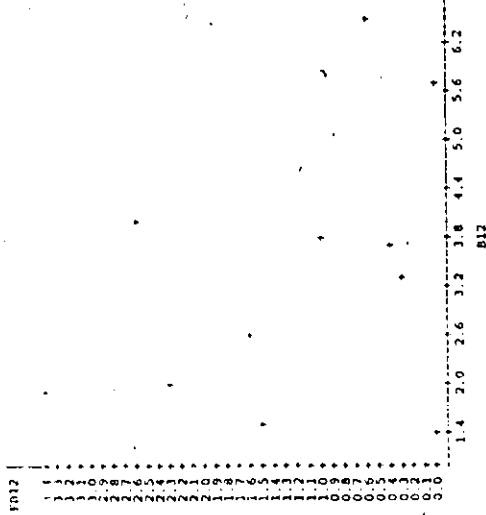
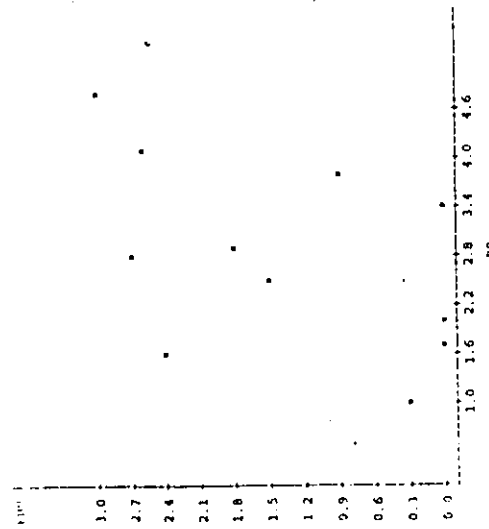
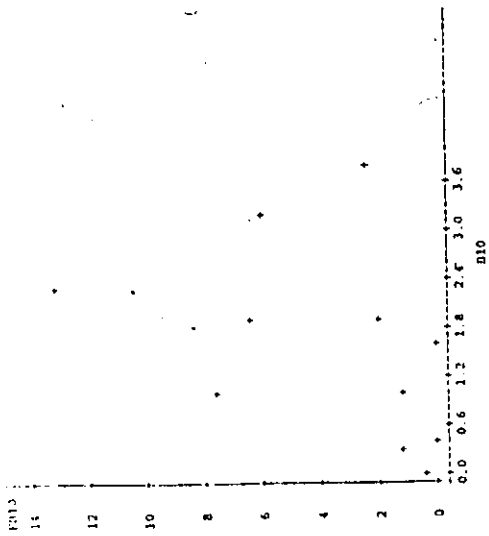


Fig. 5.2 (cont.)

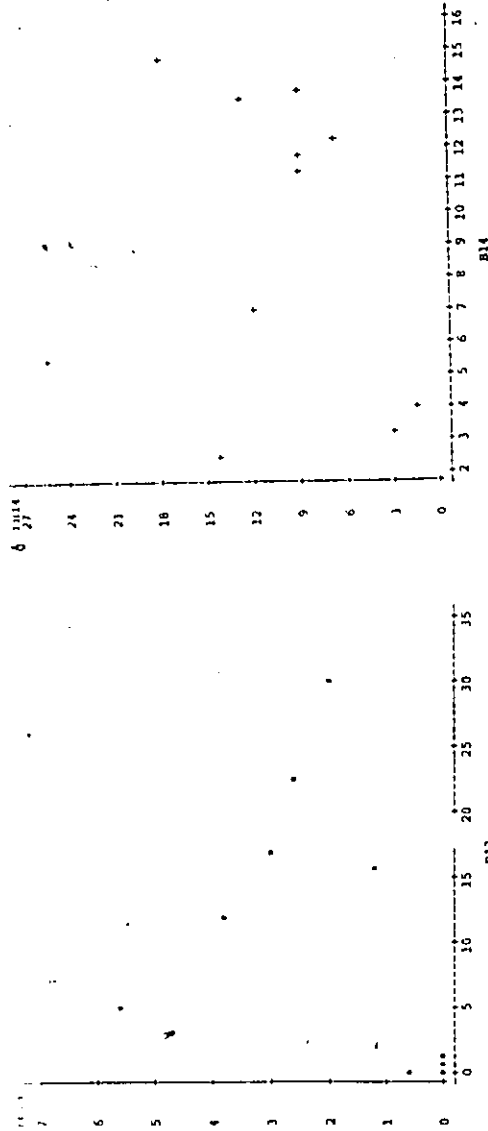


Fig. 5.2 (concluded)

## CHAPTER 6: Enlarged data set and geostatistical model

### 6.1 Introduction

The enlarged data set, ( Table 6.1 ) created by merging Stam's and Gradstein's data, is difficult to work with because of the poor correlation between the two sets. For a linear relationship the relative closeness of the relation is best measured by the coefficient of determination. This measure shows the proportion of the variance in the dependent variable that is associated with differences in the other variable. As seen in Chapter 5, the relative closeness of the relation of Stam's and Gradstein's data is weak.

### 6.2 - Tojeira 1 section (Eoguttulina sp., Epistomina mosquensis and Ophthalmidium strumosum)

The new data set has  $N = 41$  ( Stam's 31 samples plus 17 of Gradstein's sample equals 48 samples, but 7 samples were collected at the same level by both authors.). About 10 autocorrelations is a safe number to examine a taxon using an ARIMA procedure to estimate the a.c.f and p.a.c.f for Tojeira 1.

As seen in Appendix I Eoguttulina sp., Epistomina mosquensis and Ophthalmidium strumosum show an exponential function. The estimated a.c.f and p.a.c.f of Eoguttulina sp., E. mosquensis and O. strumosum suggest two things: 1 - The data of these species have a stationary mean, and 2 - The AR ( 1 ) model for Eoguttulina sp.

TABLE 6.1 ENLARGED DATA SET

SAMPLE	A1	A2	A3	A4	A5	A6	A7	A8	A9	A10	A11	A12	A13	A14
S6.2	5.0	4.1	0.0	0.0	1.5	7.9	6.6	4.8	10.4	0.0	0.0	1.8	0.9	45.3
26.3	2.5	14.8	0.0	0.0	0.0	24.7	5.7	4.9	0.2	0.0	0.0	0.9	18.8	17.9
S6.3A	0.5	19.0	0.0	0.5	0.0	22.5	3.8	4.6	1.1	0.0	0.0	0.0	26.8	13.6
S6.4	5.0	5.4	0.2	1.2	0.0	19.8	10.6	3.4	4.2	0.0	0.0	2.4	0.4	26.1
S6.5	2.6	1.8	15.1	0.0	13.3	4.7	4.9	0.0	0.0	0.0	0.0	2.9	7.2	34.8
S6.6A	1.2	13.5	8.4	1.5	16.2	5.4	0.3	0.6	14.4	0.0	0.0	13.8	6.0	11.7
S6.6	2.3	2.2	24.2	11.8	0.0	9.9	4.8	2.5	3.8	0.0	0.0	10.8	3.8	9.1
26.7	5.4	3.4	18.7	3.4	0.0	11.1	3.1	3.8	3.8	0.2	0.0	6.9	0.6	29.2
S6.8	4.5	0.3	12.4	1.0	0.0	5.6	2.1	2.0	35.4	0.0	5.1	5.3	0.0	15.4
S6.9	6.4	2.6	26.1	4.3	0.3	4.6	4.1	2.6	3.7	0.0	10.9	5.7	0.0	23.2
T1/6	7.7	0.5	27.4	5.0	0.0	13.9	1.2	1.0	4.0	1.2	14.4	5.0	3.5	14.2
S6.10	5.1	1.1	27.6	2.5	2.5	1.4	0.6	2.2	1.1	0.3	29.6	1.7	0.3	20.7
T1/7	2.4	0.7	38.5	23.2	0.0	15.2	2.2	0.5	2.7	2.9	6.4	1.0	0.5	2.2
S6.11	3.7	1.1	13.2	17.6	0.0	23.0	6.1	2.0	0.5	0.2	10.9	1.1	0.2	11.1
T1/8	5.0	0.6	25.1	4.7	0.0	24.2	2.4	2.9	4.1	1.2	9.4	3.2	2.6	11.8
S6.12	21.6	1.3	23.6	3.0	3.0	9.3	3.3	0.0	0.0	0.0	12.3	1.3	0.0	15.0
T1/9	32.1	0.0	22.4	8.9	3.0	8.2	2.2	2.2	0.0	0.0	11.2	1.5	1.5	6.7
S6.13	3.9	0.6	23.3	5.7	4.8	8.7	2.7	0.3	1.2	0.0	14.0	1.5	0.0	23.0
S6.14	7.6	1.5	15.5	0.7	0.9	4.8	3.8	1.8	2.4	2.3	4.8	5.0	0.7	31.1
T1/10	7.8	0.4	18.8	15.3	0.4	11.8	1.6	2.7	7.8	0.0	7.8	4.7	5.1	15.7
S6.15	2.4	0.3	35.2	7.8	6.0	8.1	7.2	1.5	0.6	0.3	10.5	1.5	0.0	15.1
T1/11	10.6	1.2	32.4	1.9	0.0	14.8	1.6	3.9	4.7	5.9	5.9	3.5	1.9	9.4
S6.16	7.8	0.5	21.4	3.1	7.7	15.2	5.5	1.4	2.1	2.9	15.0	1.7	0.0	11.2
T1/12	9.4	1.0	25.7	2.0	0.0	7.4	1.7	3.0	6.4	5.7	3.5	2.5	3.2	26.5
S6.17	2.9	1.1	46.3	2.0	1.9	5.1	2.8	1.1	1.4	8.3	6.4	4.1	0.0	13.2
S6.18	14.0	0.3	18.0	1.7	5.3	3.9	7.2	2.8	0.8	6.6	8.0	4.7	0.0	20.8
S6.19	8.5	0.8	39.7	0.6	3.5	3.5	4.4	2.5	1.0	1.9	11.6	4.1	0.0	8.3
T1/13	17.9	0.0	9.2	3.0	0.8	19.8	2.7	3.8	4.9	10.3	3.0	4.2	0.0	19.1
S6.20	6.0	0.8	30.4	6.0	1.4	9.1	7.4	1.5	1.9	0.4	12.6	4.4	1.0	9.2
S6.21	11.4	0.8	13.5	1.7	1.2	15.0	11.3	1.9	1.9	7.2	9.3	4.2	0.0	13.5
S6.22	4.6	3.5	23.0	0.3	0.8	11.8	6.6	1.6	0.5	2.4	11.8	7.6	0.0	22.3
S6.23	3.3	1.5	33.5	0.8	6.2	11.4	9.3	2.1	1.6	1.9	10.9	4.8	0.0	7.7
S6.24	7.1	0.6	37.1	3.2	3.1	6.6	3.6	1.5	1.2	5.8	1.0	3.4	0.0	9.0
S6.25	7.4	0.6	36.6	7.4	8.4	9.1	1.2	3.9	0.6	1.9	0.0	1.9	0.0	13.6
S6.26	17.1	0.9	38.8	4.9	0.6	6.7	2.7	1.7	0.6	2.9	3.2	0.6	0.0	13.3
S6.26A	2.2	1.0	34.8	3.7	3.4	16.2	3.3	2.5	0.3	5.2	9.1	4.6	0.0	7.0
T1/16	3.3	0.0	39.0	1.9	0.0	25.7	2.4	0.9	4.8	2.8	1.9	2.4	2.9	10.5
T1/17	3.3	0.0	38.9	12.9	0.7	16.5	2.6	1.0	0.7	1.3	15.5	0.3	0.0	5.3
S6.27	10.6	0.3	36.5	1.2	1.8	11.9	8.8	0.3	1.2	9.7	6.4	5.5	0.0	8.8
S6.28	11.8	0.0	27.4	3.4	2.3	14.0	2.4	2.0	0.3	1.1	17.1	2.3	0.0	9.4
S6.29	6.2	0.2	16.7	3.2	2.5	6.5	0.6	2.5	0.0	5.3	21.6	4.9	0.4	16.0

( Fig. 6.1 ), AR( 2 ) model for E. mosquensis ( Fig. 6.2), and ARMA (1, 1) model for O. strumosum ( Fig. 6.3) are good first choices to try at the estimation stage. The stationarity assumption is confirmed since  $\hat{\phi}_1 = 0.510$ ,  $\hat{\phi}_1 = 0.281$  and  $\hat{\phi}_1 = 0.935$ , respectively for the mentioned species. These satisfy the condition  $\hat{\phi}_1 < 1$ . The large t - values attached to  $\hat{\phi}_1$  indicate that these terms should be kept in the model.

As seen in Figs. 6.1- 6.3 none of the residual autocorrelations have an absolute  $t_b$  - value greater than the practical warning levels. The chi-squared values indicate this too.

The constant c and a values were estimated from their best - fitting straight lines as shown in Figs. 6.4 - 6.6. Table 6.2 shows the estimated values of a and c.

TABLE 6.2 Estimated a and c values for best - fitting lines for A2 ( Eoguttulina sp. ), A3 ( Epistomina mosquensis ) and A10 ( Ophthalmidium strumosum ) for N = 41

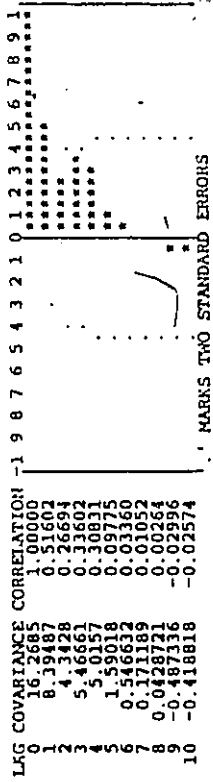
Taxa	A2( <u>Eoguttulina</u> sp.)	A3( <u>E. mosquensis</u> )	A10( <u>O. strumosum</u> )
c	0.48	0.52	0.60
a	0.08	0.12	0.19

Eq. 56 and Eq. 60 were used to calculate  $F_N$  and  $F_L$  ( Tables 6.3 - 5 ) respectively. The theoretical and observed relative variances were plotted against the lag as seen in Fig. 6.7 - 6.9 for each species. The correspondence between the experimental and theoretical values are not good for these taxa. As explained above, Stam's and Gradstein's data do

Figure 6.1 Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(1) time series model to A2 (Eoguttulina sp.) with N = 41

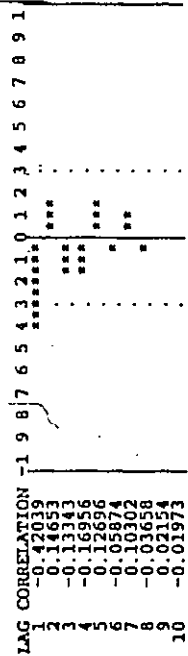
SAS 15:44 TUESDAY, JULY 28, 1987  
 ARIHA PROCEDURE  
 ARIHA: MAXIMUM LIKELIHOOD ESTIMATION

NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES = 2.20244  
 STANDARD DEVIATION = 4.03343  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

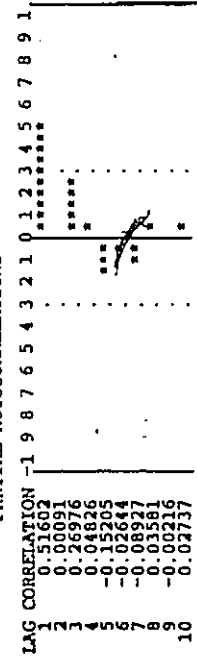


MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS



PARTIAL AUTOCORRELATIONS



AUTOCORRELATION CHECK FOR WHITE NOISE

TO CHI SQUARE DF PROB AUTOCORRELATIONS  
 LAG 6 25.25 6 0.000 0.516 0.267 0.336 0.308 0.098 0.034

SAS 15:44 TUESDAY, JULY 28, 1987  
 ARIHA: MAXIMUM LIKELIHOOD ESTIMATION

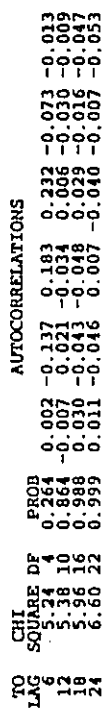
PARAMETER ESTIMATE STD. ERROR T RATIO LAG  
 MU 2.19987 1.09108 2.01 0  
 AR1,1 0.509558 0.13672 3.73 1

CONSTANT ESTIMATE = 1.07891  
 VARIANCE ESTIMATE = 12.4988  
 STD ERROR ESTIMATE = 3.53536  
 AIC = 222.154  
 SBC = 225.581  
 NUMBER OF RESIDUALS =

CORRELATIONS OF THE ESTIMATES

MU 1.000 AR1,1  
 AR1,1 -0.024 1.000

AUTOCORRELATION CHECK OF RESIDUALS



MODEL FOR VARIABLE A2 ESTIMATED MEAN = 2.19987

AUTOREGRESSIVE FACTORS  
 FACTOR 1  
 1-.509558B\*\*(1)

Figure 6.2 Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the AR(2) time series model to A3(E. mosquensis) with N = 41

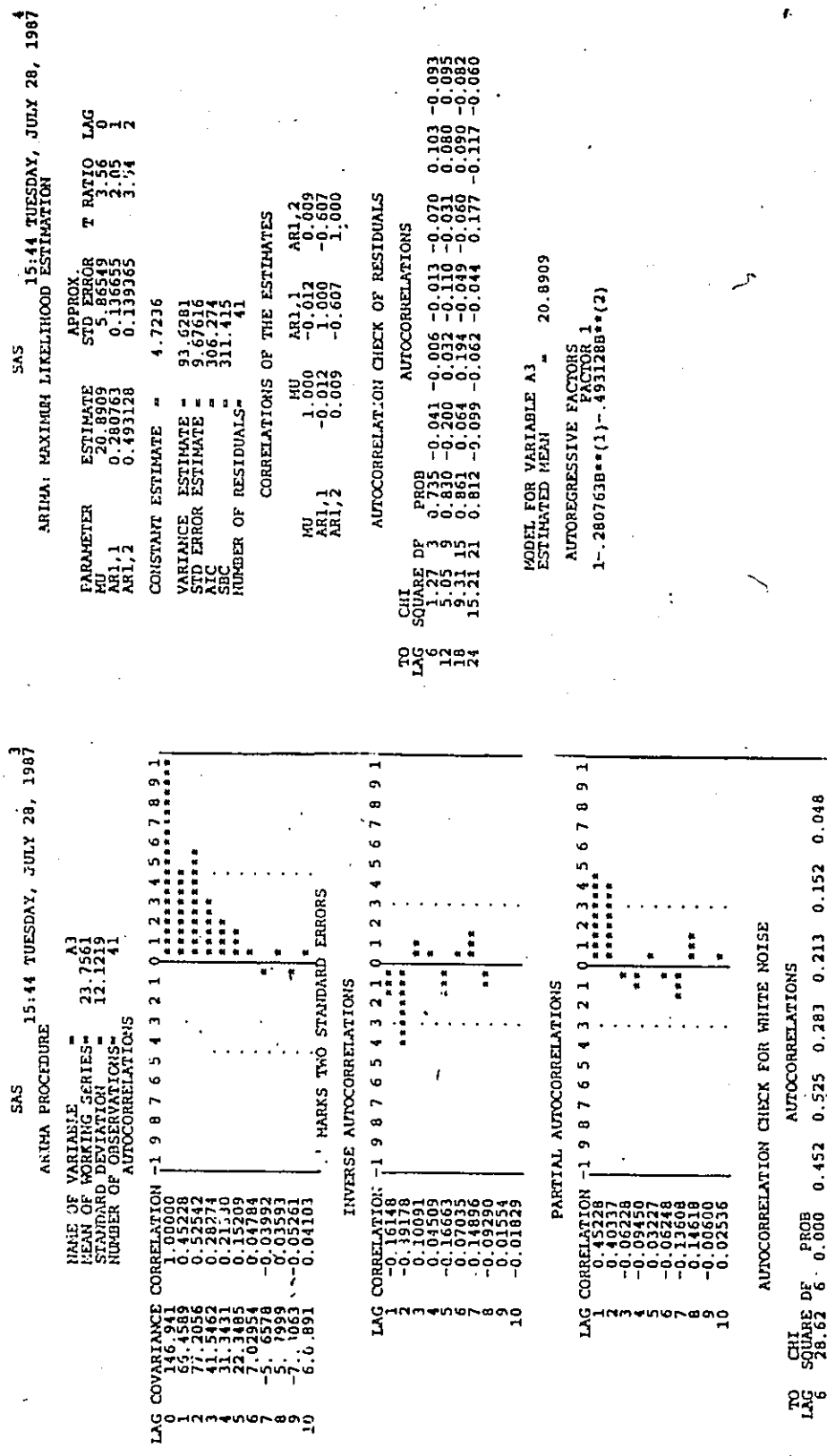


Figure 6.3 Printout of estimated a.c.f.'s and p.a.c.f.'s after fitting the ARMA(1,1) time series model to A10(O. strumosum) with N = 41

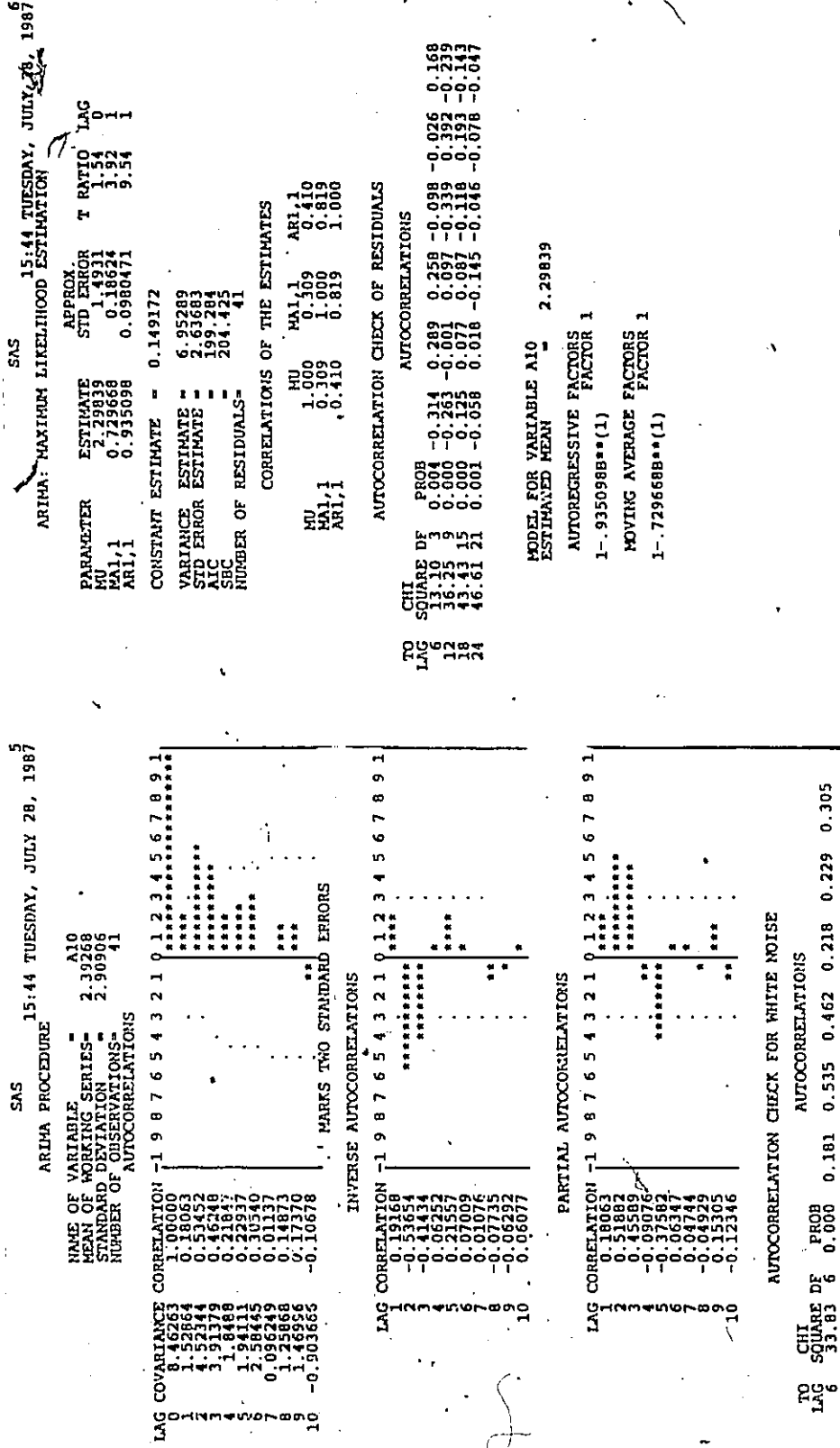
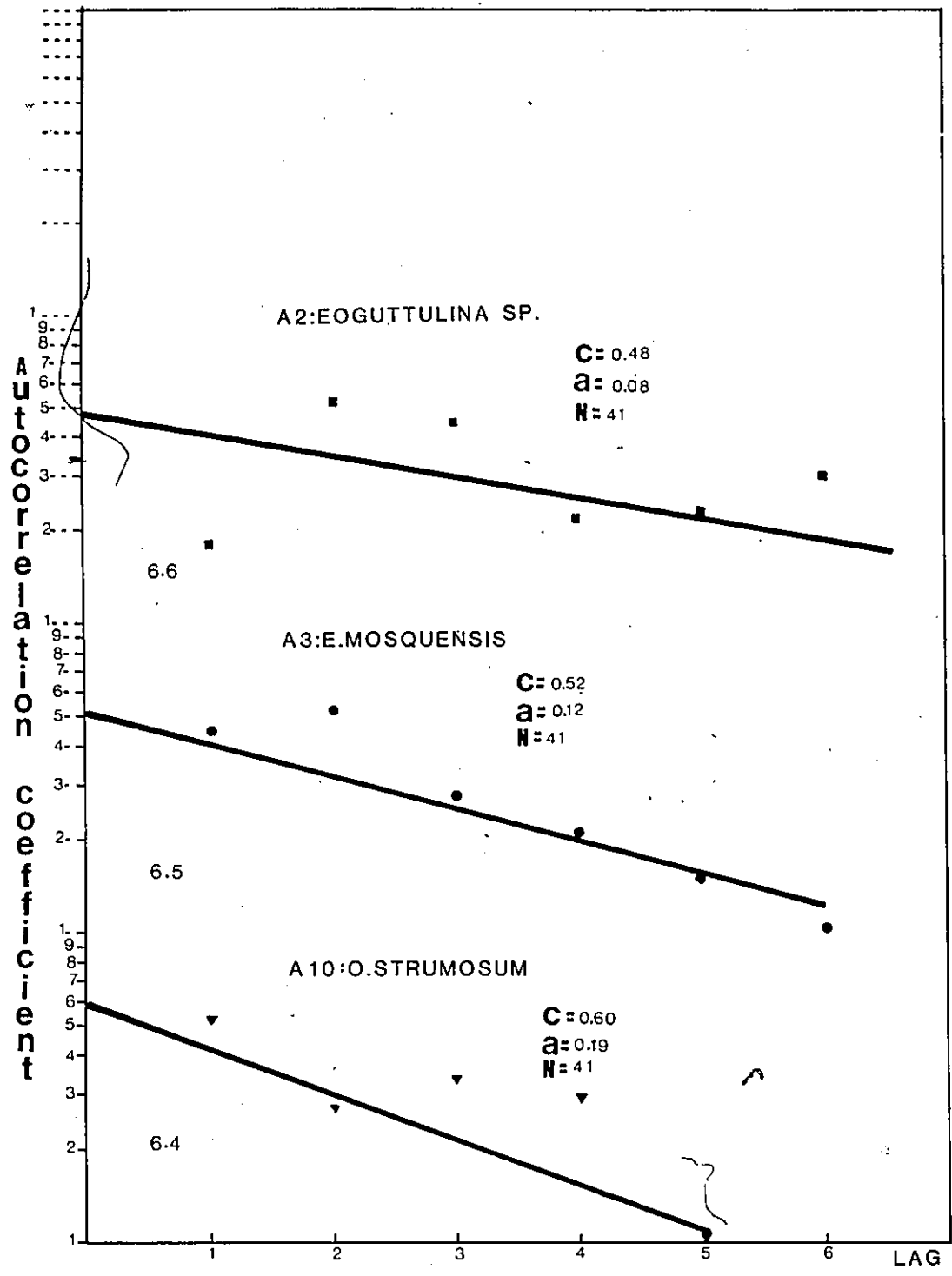


Figure 6.4 Correlogram for A10 (O. strumosum) and best - fitting exponential curve with  $a = 0.19$  and  $c = 0.60$

Figure 6.5 Correlogram for A3 (E. mosquensis) and best - fitting exponential curve with  $a = 0.12$  and  $c = 0.52$

Figure 6.6 Correlogram for A2 (Eoguttulina sp.) and best - fitting exponential curve with  $a = 0.08$  and  $c = 0.48$



not show strong linear relationships with one another.

TABLE 6.3 Theoretical and observed values of the relative variance of  
A2 (Eogiuttulina sp.)

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.52
1	16.67	1.0	0.99
2	12.91	1.55	1.43
3	10.22	1.84	1.85
4	7.78	1.87	2.25
5	6.50	1.95	2.63

TABLE 6.4 Theoretical and observed values of the relative variance of  
A3 (Epistomina mosquensis)

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.48
1	150.62	1.0	0.98
2	108.08	1.43	1.44
3	94.30	1.88	1.87
4	80.95	2.15	2.26
5	68.25	2.26	2.63
6	58.61	2.33	2.97

TABLE 6.5 Theoretical and observed values of the relative variance of A10 (Ophthalamidium strumosum )

k(lag)	$S^2(x)$	$F_N$	$F_L$
0	0	0	0.40
1	8.67	1.0	0.95
2	5.16	1.2	1.45
3	4.74	1.64	1.90
4	4.49	2.07	2.20
5	4.24	2.44	2.63
6	4.03	2.80	2.97

The probit transformed ( Tables 6.6, 6.8 and 6.10 ) frequency distributions ( Tables 6.7, 6.9 and 6.11 ) for each species were plotted on normal distribution paper for single, paired and sets of 4 consecutive values as done previously ( Fig. 6.10a, b, c - 6.11a, b, c and 6.12a, b, c.).

Eq. 60 represents a prob - normal distribution which is a straight line on probability paper. For Eoguttulina sp., Epistomina mosquensis and Ophthalamidium strumosum , the prob - normal distribution parameters, which are  $x$ ,  $s^2(x)$ ,  $R$  and  $b$  values, were calculated as shown in Tables 6.12 - 14. Calculated parameters were used with Eq. 60 and  $Y$  functions were plotted on normal probability paper as theoretical distributions as shown in Fig. 6.10a, b, c - 6.11a, b, c and 6.12a, b,c.

# THEORETICAL CURVE

A2:EOGUTTULINA SP.

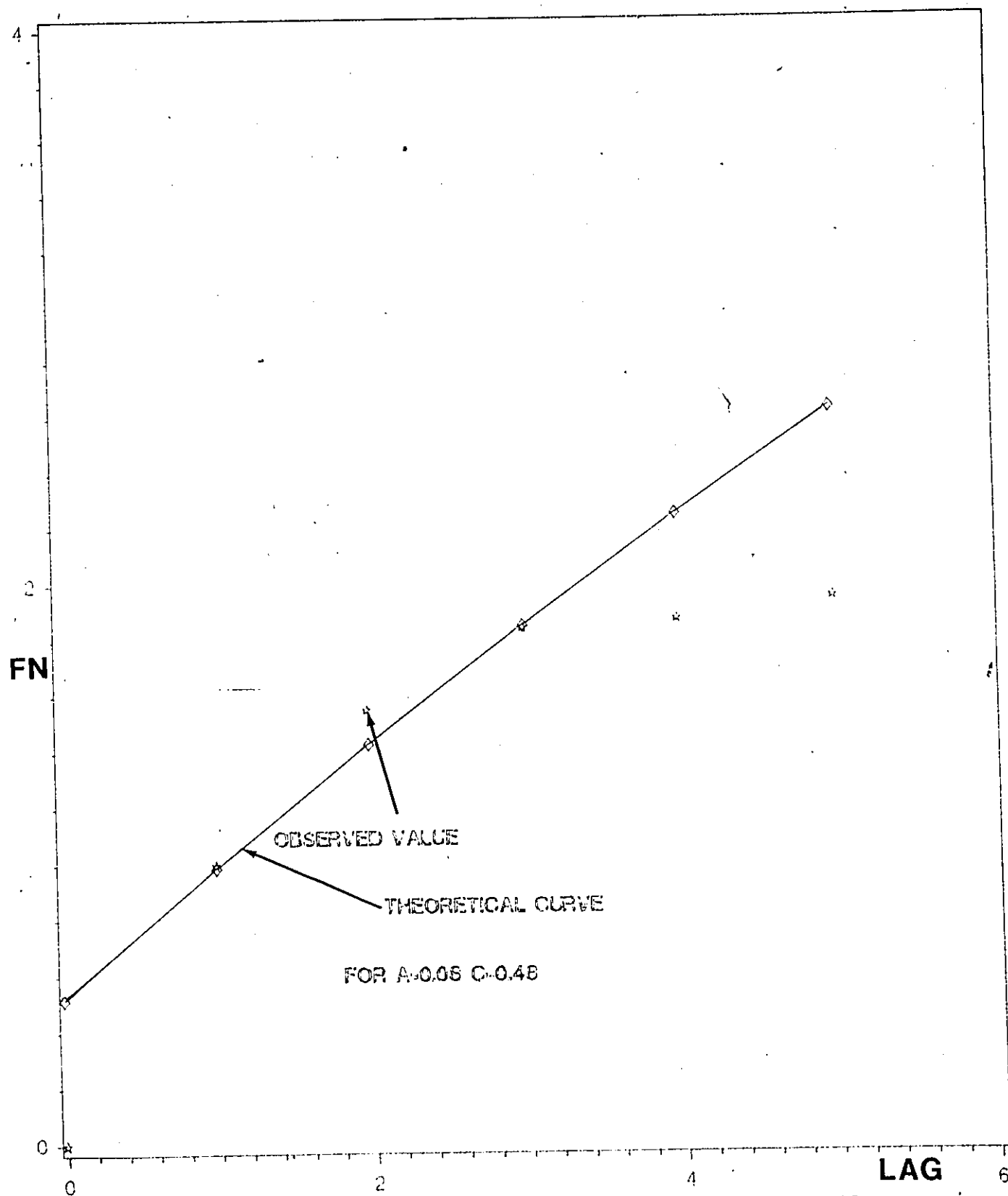


Figure 6.7 Theoretical curve for  $a = 0.08$  and  $c = 0.48$  in comparison with observed values of A2 (Eoguttulina sp.)

# THEORETICAL CURVE

A3:EMOSQUENSIS

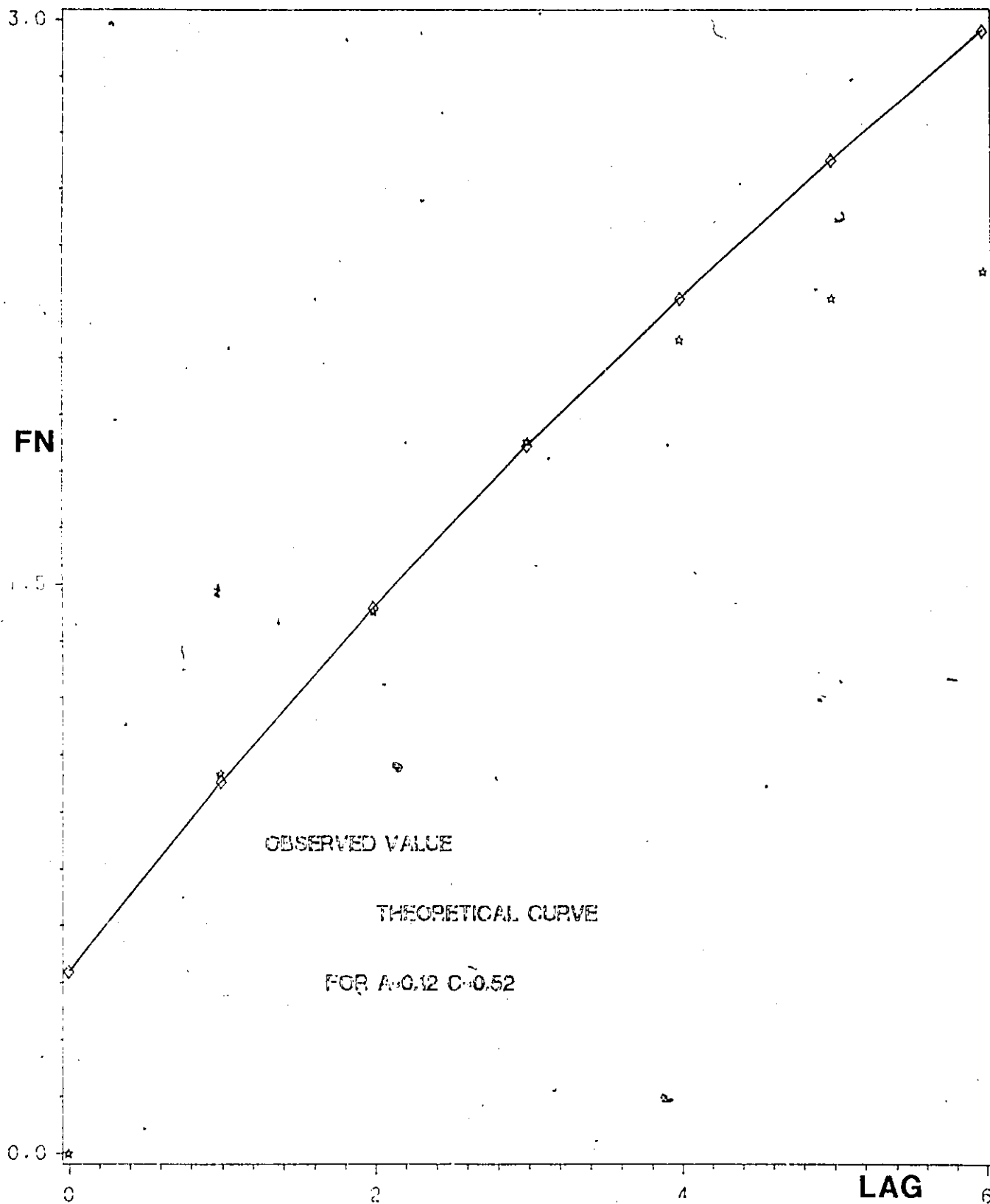


Figure 6.8 Theoretical curve for a 0.12 and c 0.52 in comparison with observed values of E. mosquensis

## THEORETICAL CURVE

A10:O.STRUMOSUM

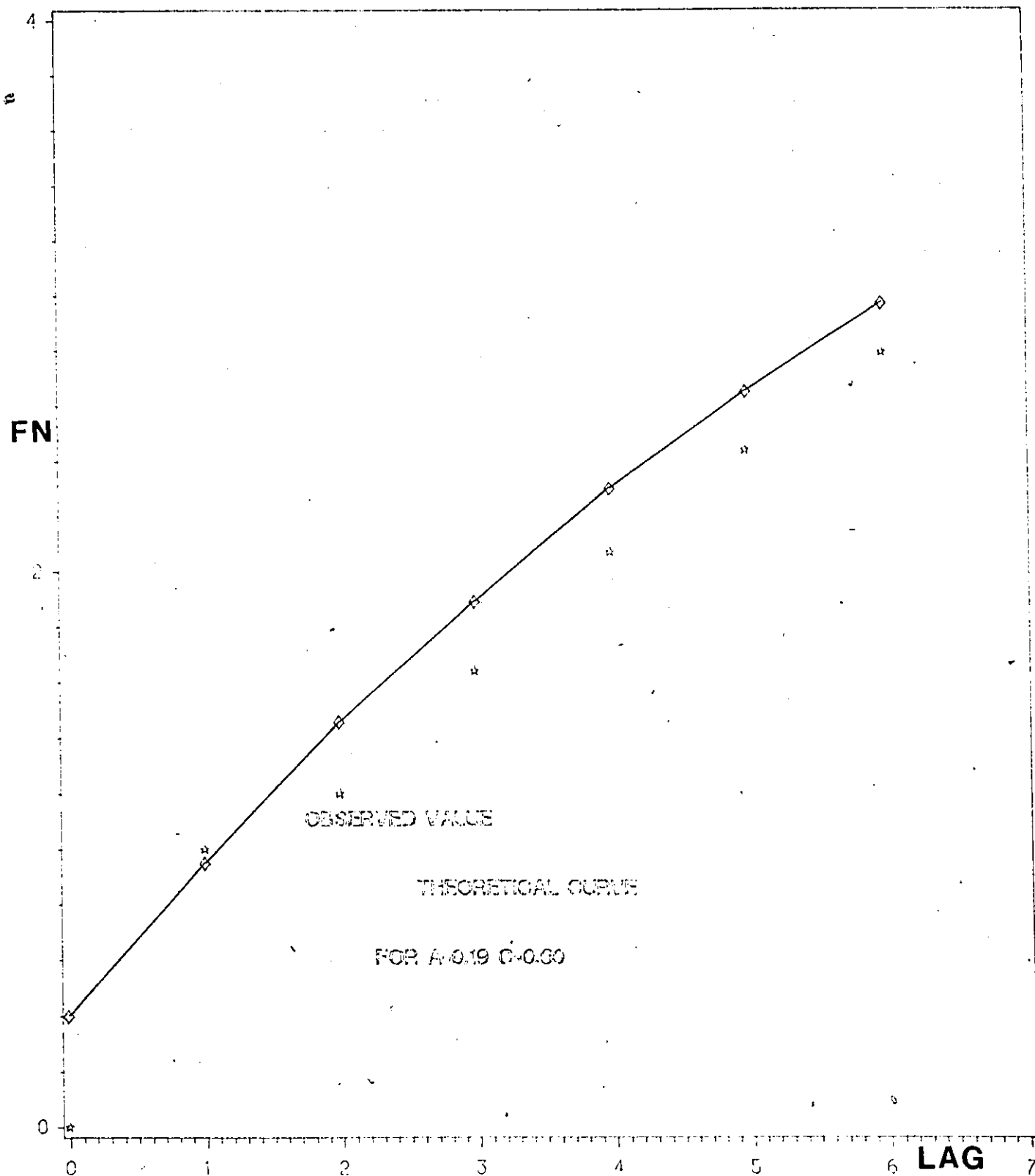


Figure 6.9 Theoretical curve for a 0.19 and c 0.60  
in comparison with observed values of *O. strumosum*

Table 6.6 probit transformation of A2 ( Eoguttulina sp. ) for N = 41, 20 and 10 samples from the Tojeira 1 section.

sample single	probit single	sample pairs	probit pairs	sample set of 4	probit set of 4
4.1	3.261				
14.8	3.955	9.45	3.686		
19.0	4.122				
5.4	3.393	12.2	3.835	10.825	3.764
1.8	2.903				
13.5	3.897	7.65	3.570		
2.2	2.986				
3.4	3.175	2.80	3.080	5.225	3.376
0.3	2.252				
2.6	3.057	1.45	2.816		
0.5	2.424				
1.1	2.710	0.80	2.591	1.125	2.766
0.7	2.543				
1.1	2.710	0.90	2.634		
0.6	2.488				
1.3	2.774	0.45	2.386	0.675	2.644
0.0	2.187				
0.6	2.488	0.30	2.252		
1.5	2.830				

( continued )

Table 6.6 (concluded)

0.4	2.348	0.85	2.612	0.575	2.475
0.3	2.252				
1.2	2.743	0.75	2.567		
0.5	2.424				
1.0	2.674	0.75	2.567	0.750	2.567
1.1	2.710				
0.3	2.252	0.70	2.543		
0.8	2.591				
0.0	2.187	0.40	2.348	0.550	2.456
0.8	2.591				
0.8	2.591	0.8	2.591		
3.5	3.188				
1.5	2.830	2.50	3.040	1.650	2.868
0.6	2.488				
0.6	2.483	0.60	2.488		
0.9	2.634				
1.0	2.674	0.95	2.654	0.775	2.579
0.0	2.187				
0.0	2.187	0.0	2.187		
0.3	2.252				
0.0	2.187	0.15	2.187	0.00	2.187
0.2	2.187				

Table 6.7 Cumulative frequency distributions of Eoguttulina sp. for N = 41, 20 and 10 samples from the Tojeira 1 section.

N = 41

Class interval	Freq.	cum. freq	perc. %	cum. %	plotting value
2.25 - 2.55	18	18	43.90	43.90	42.74
2.55 - 2.85	13	31	31.71	75.61	74.19
2.85 - 3.15	3	34	7.32	82.93	81.45
3.15 - 3.45	4	38	9.76	92.68	91.13
3.45 - 3.75	0	38	0.00	92.68	91.13
3.75 - 4.05	3	41	7.32	100.00	98.40

N = 20

2.2 - 2.6	11	11	55.00	55.00	52.50
2.6 - 3.0	4	15	20.00	75.00	72.13
3.0 - 3.4	2	17	10.00	85.00	82.00
3.4 - 3.8	2	19	10.00	95.00	91.80
3.8 - 4.2	1	20	5.00	100.00	96.72

N = 10

2.2 - 2.6	5	5	50.00	50.00	45.16
2.6 - 3.0	3	8	30.00	80.00	74.19
3.0 - 3.4	1	9	10.00	90.00	83.87
3.4 - 3.8	1	10	10.00	100.00	93.55

Table 6.8 Probit transformation of A3 ( Epistomina mosquensis ) for  
 N = 41,20 and 10 samples from the Tojeira 1 section.

sample single	probit single	sample pairs	probit pairs	sample set of 4	probit set of 4
0.0	2.187				
0.0	2.131	0.00	2.187		
0.0	2.131				
0.2	2.187	0.00	2.187	0.00	2.187
15.1	3.968				
8.4	3.621	11.75	3.812		
24.2	4.300				
18.1	4.081	21.15	4.198	16.45	4.024
12.4	3.845				
26.1	4.360	19.25	4.131		
27.4	4.399				
27.6	4.405	27.5	4.402	23.375	4.273
38.5	4.708				
13.2	3.883	25.85	4.357		
25.1	4.329				
23.6	4.281	24.35	4.304	25.10	4.329
22.4	4.241				
23.3	4.271	22.85	4.258		
15.5	3.985				

( continued )

Table 6.8 ( concluded )

18.8	4.115	17.15	4.047	20.0	4.158
35.2	4.620				
32.4	4.543	33.8	4.582		
21.4	4.207				
25.7	4.347	23.55	4.280	28.675	4.437
46.3	4.907				
18.0	4.085	32.15	4.535		
39.7	4.739				
9.2	3.671	24.45	4.308	28.30	4.426
30.4	4.487				
13.5	3.897	21.95	4.226		
23.0	4.261				
33.5	4.574	28.25	4.424	25.10	4.329
37.1	4.671				
36.6	4.658	36.85	4.638		
38.8	4.715				
34.8	4.609	36.8	4.636	36.825	4.463
39.0	4.721				
38.9	4.718	38.95	4.718		
36.5	4.655				
27.4	4.399	31.95	4.532	35.45	4.626
16.7	4.34				

Table 6.9 Cumulative frequency distribution of A3( Epistomina mosquensis )  
for N = 41, 20 and 10 samples for Tojeira 1 section

N = 41

class interval	freq.	cum. freq.	perc. %	cum. %	plotting value
2.25 - 2.75	4	4	9.76	9.76	8.87
2.75 - 3.25	0	4	0.00	9.76	8.87
3.25 - 3.75	2	6	4.88	14.63	13.70
3.75 - 4.25	11	17	26.83	41.46	40.32
4.25 - 4.75	23	40	56.10	97.56	95.96
4.75 - 5.25	1	41	2.44	100.00	98.40

N = 20

2.25 - 2.75	3	3	15.00	15.00	13.11
2.75 - 3.25	0	3	0.00	15.00	13.11
3.25 - 3.75	0	3	0.00	15.00	13.11
3.75 - 4.25	4	7	20.00	35.00	32.80
4.25 - 4.75	13	20	65.00	100.00	96.72

N = 10

2.1 - 2.7	1	1	10.00	10.00	6.45
2.7 - 3.3	0	1	0.00	10.00	6.45
3.3 - 3.9	0	1	0.00	10.00	6.45
3.9 - 4.5	9	10	90.00	100.00	93.55

Table 6.10 Probit transformation of A10 (*Ophthalidium strumosum*)  
for N = 41, 20 and 10 samples from Tojeira 1 section.

sample single	probit single	sample pairs	probit pairs	sample set of 4	probit set of 4
0.0	2.187				
0.0	2.187	0.00	2.187		
0.0	2.187				
0.0	2.187	0.00	2.187	0.00	2.187
0.0	2.187				
0.0	2.187	0.00	2.187		
0.0	2.187				
0.2	2.187	0.00	2.187	0.00	2.187
0.0	2.187				
0.0	2.187	0.00	2.187		
1.2	2.187				
0.3	2.252	0.75	2.567	0.375	2.324
2.9	3.104				
0.2	2.187	1.55	2.843		
1.2	2.187				
0.0	2.187	0.60	2.488	1.075	2.701
0.0	2.187				
0.0	2.187	0.00	2.781		
2.3	3.005				

( continued )

Table 6.10 (concluded)

0.0	2.187	1.15	2.715	0.575	2.472
0.3	2.252				
5.9	3.437	3.10	3.134		
2.9	3.104				
5.7	3.420	4.30	3.283	3.70	3.213
8.3	3.615				
6.6	3.494	7.45	3.556		
1.9	2.925				
0.3	2.252	1.10	2.710	4.275	3.280
0.4	2.348				
7.2	3.539	3.80	3.226		
2.4	3.023				
1.9	2.925	12.15	2.969	2.975	3.115
5.8	3.428				
1.9	2.925	3.85	3.236		
2.9	3.104				
5.2	3.374	4.05	3.255	3.95	3.243
2.8	3.089				
1.3	2.774	2.05	2.956		
9.7	3.701				
1.1	2.710	5.40	3.393	3.725	3.216
5.3	3.384				

Table 6.11 Cumulative frequency distributions of Ophthalmidium strumosum  
for N = 41, 20 and 10 samples

N = 41

class interval	freq.	cum. freq.	perc. %	cum. %	plotting value
2.125 - 2.375	19	19	46.34	46.34	45.16
2.375 - 2.625	0	19	0.00	46.34	45.16
2.625 - 2.875	4	23	9.76	56.10	54.84
2.875 - 3.125	9	32	21.95	78.05	76.60
3.125 - 3.375	1	33	2.44	80.49	79.00
3.375 - 3.625	8	41	19.51	100.00	98.40

N = 20

2.25 - 2.55	7	7	35.00	35.00	32.80
2.55 - 2.85	4	11	20.00	55.00	52.46
2.85 - 3.15	3	14	15.00	70.00	67.21
3.15 - 3.45	5	19	25.00	95.00	91.80
3.45 - 3.75	1	20	5.00	100.00	96.72

N = 10

2.25 - 2.55	4	4	40.00	40.00	35.50
2.55 - 2.85	1	5	10.00	50.00	45.20
2.85 - 3.15	1	6	10.00	60.00	54.80
3.15 - 3.45	4	10	40.00	100.00	93.55

TABLE 6.12 Means, variances, R and b parameters for species A2, A3 and A10 for N=41

Taxon	<u>Eoguttulina sp.</u>	<u>E.mosquensis</u>	<u>O. Strumosum</u>
$\bar{x}$	0.022025	0.23756	0.023927
$s^2(x)$	0.001627	0.014694	0.000846
b	2.014	0.714	1.977
R	0.60	0.39	0.47
N	41	41	41

TABLE 6.13 Means, variances, R and b parameters for species A2, A3 and A10 for N=20

Taxa	<u>Eoguttulina sp.</u>	<u>E.mosquensis</u>	<u>O.strumosum</u>
$\bar{x}$	0.022025	0.23756	0.023927
$s^2(x)$	0.001277	0.011020	0.0006347
b	2.014	0.714	1.977
R	0.55	0.34	0.42
N	20	20	20

TABLE 6.14 Means, variances, R and b parameters for species A2, A3 and A10 for N=10

Taxa	<u>Eoguttulina sp.</u>	<u>E.mosquensis</u>	<u>O.strumosum</u>
$\bar{x}$	0.022025	0.23756	0.023927
$s^2(x)$	0.001037	0.008816	0.000487
b	2.014	0.714	1.977
R	0.51	0.29	0.38
N	10	10	10

Figure 6.10a,b,c Experimental frequency distributions for 41, 20 and 10 values of Eoguttulina-sp. plotted on two normal probability scales. The values of  $b$  and  $R$  were obtained from the means and variances. Theoretical distributions are the probnormal distributions which satisfy Eq.60

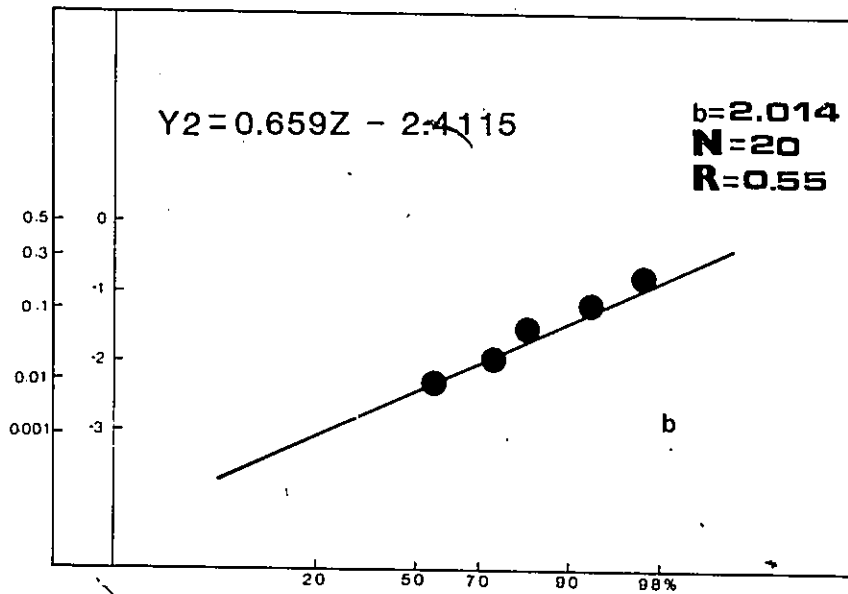
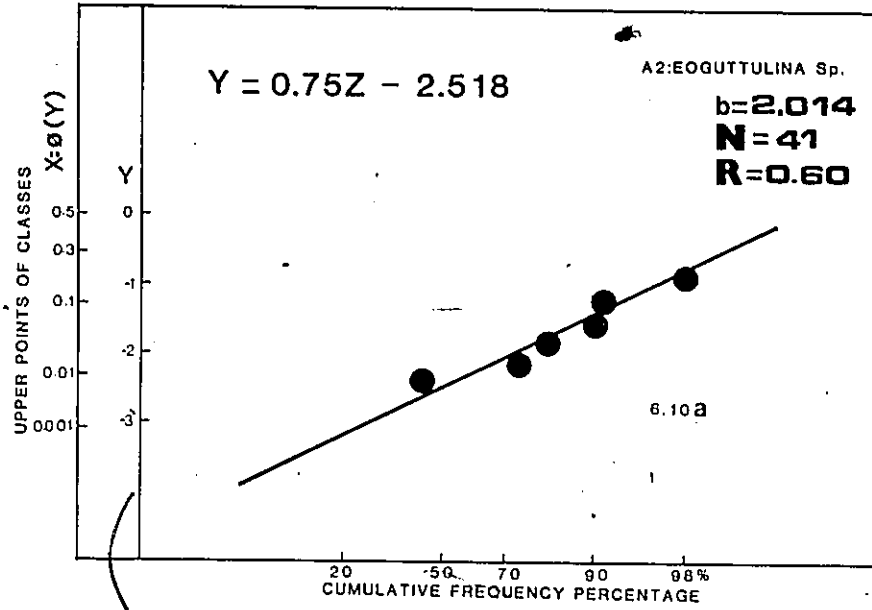
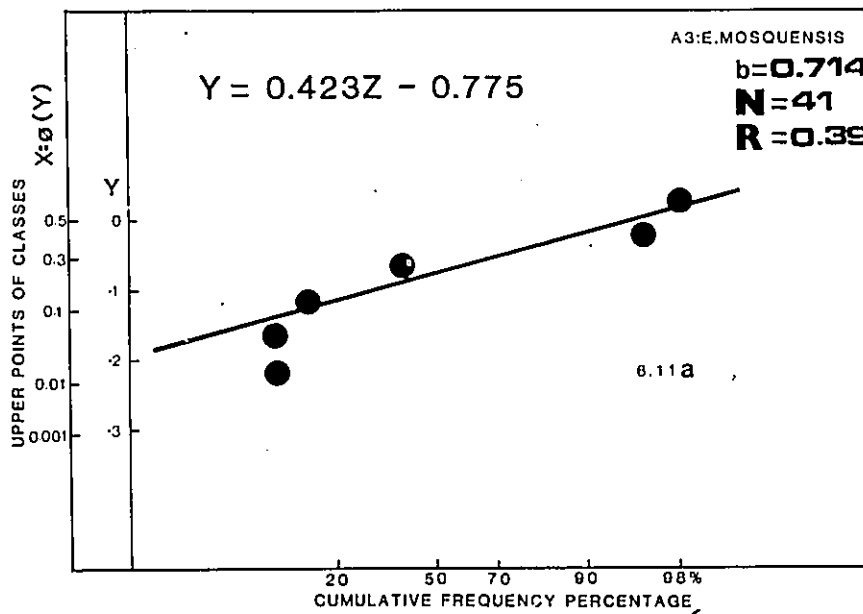
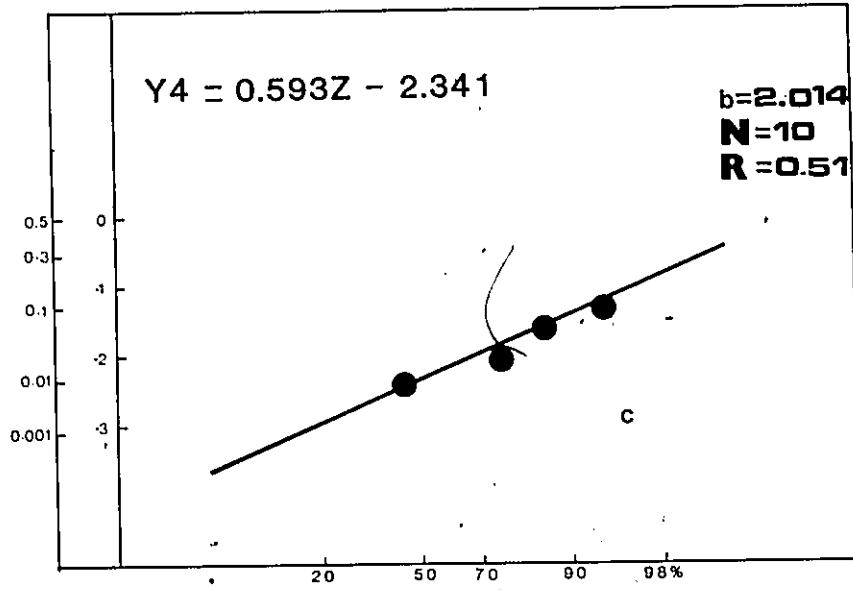


Figure 6.11a,b,c Experimental and theoretical frequency distributions for 41 (single values), 20 (pairs) and 10 (sets of four adjacent samples) for E. mosquensis



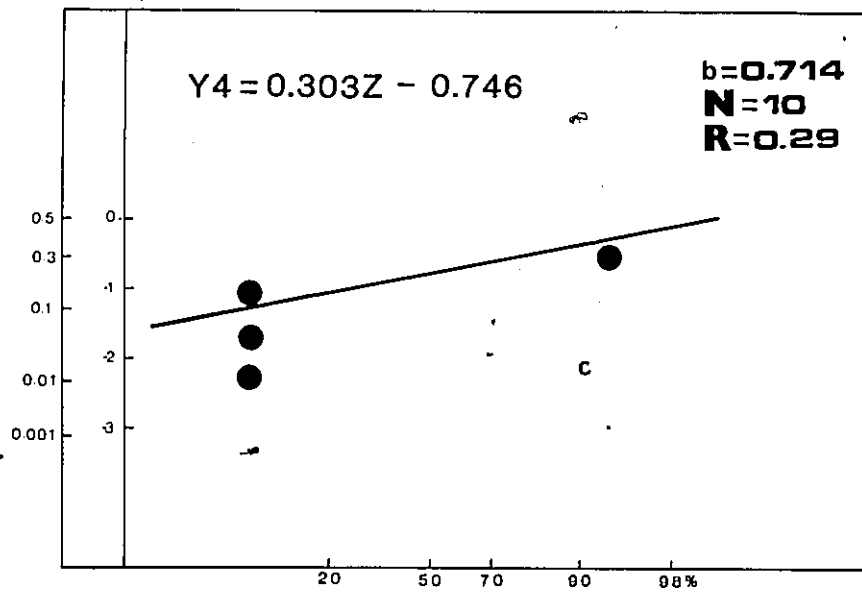
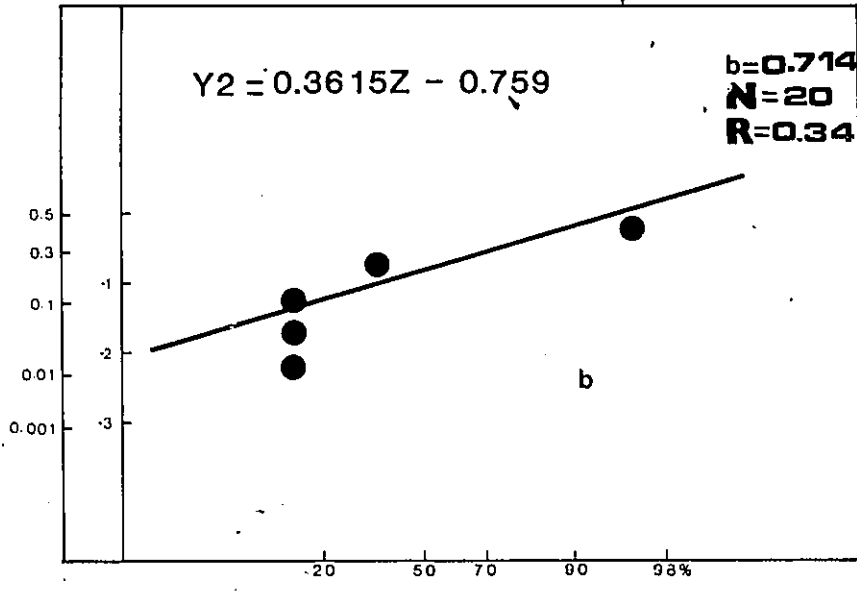
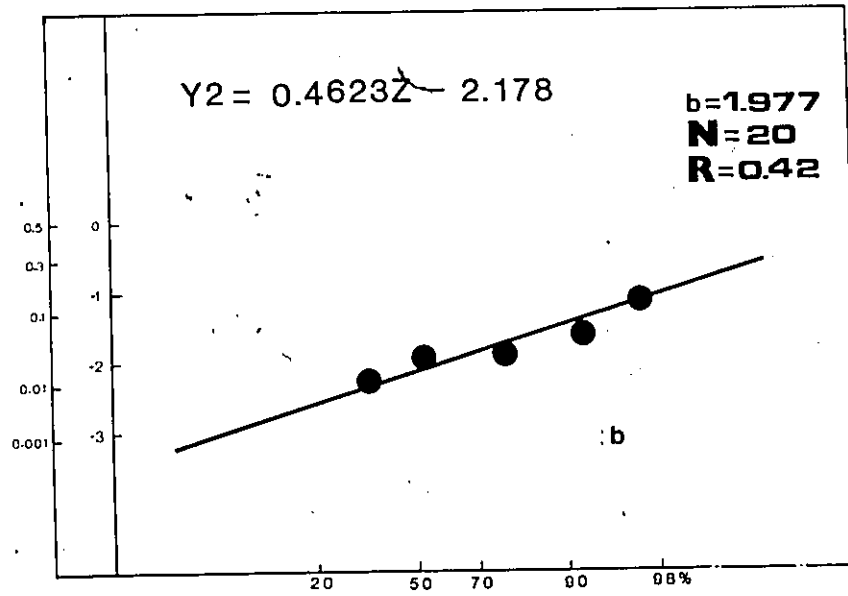
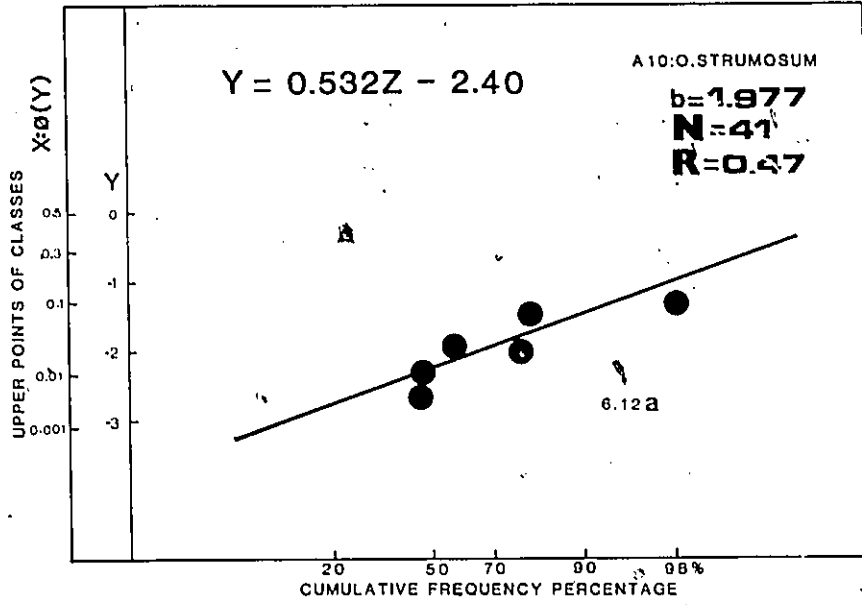
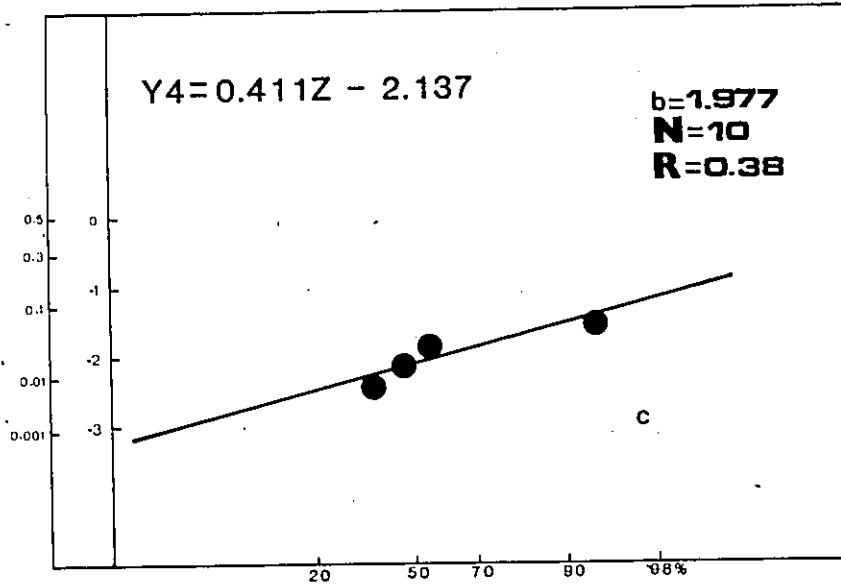


Figure 6.12a,b,c Experimental and theoretical distributions for 41, 20 and 10 values of O. strumosum. They were plotted on probability paper as explained in the text







Observed frequency distributions of Eoguttulina sp. and Epistomina mosquensis have a good fit with the theoretical probability normal distributions. The observed frequency of Ophthalmidium strumosum a good fit with its theoretical distributions, but not as good as those of Epistomina mosquensis and Eoguttulina sp.

### 6.3 Tojeira 2 section

Gradstein's samples matched Stam's samples at many levels. Just 2 of Gradstein's samples were collected from different levels than Stam's. This results in a sample number of 32 for the enlarged data set of Tojeira 2. This newly created data set is not much different from the old data set of Tojeira 2. For this reason the enlarged data set was not studied by author.

Five samples were collected in the same layer at the Tojeira 1 section. Each sample contains the same percent of different species. If the percent of species were plotted against sample numbers, the lines should be horizontal. Actually such a plot shows that they are not horizontal or that they are not even close to straight lines as shown in Fig. 6.13. The reason for this may be that each group of fossils presents counting problems or some species were grouped together as one species. In addition, during the sampling process, some species could be lost.

### 6.4 Binomial distribution and standard deviation of noise

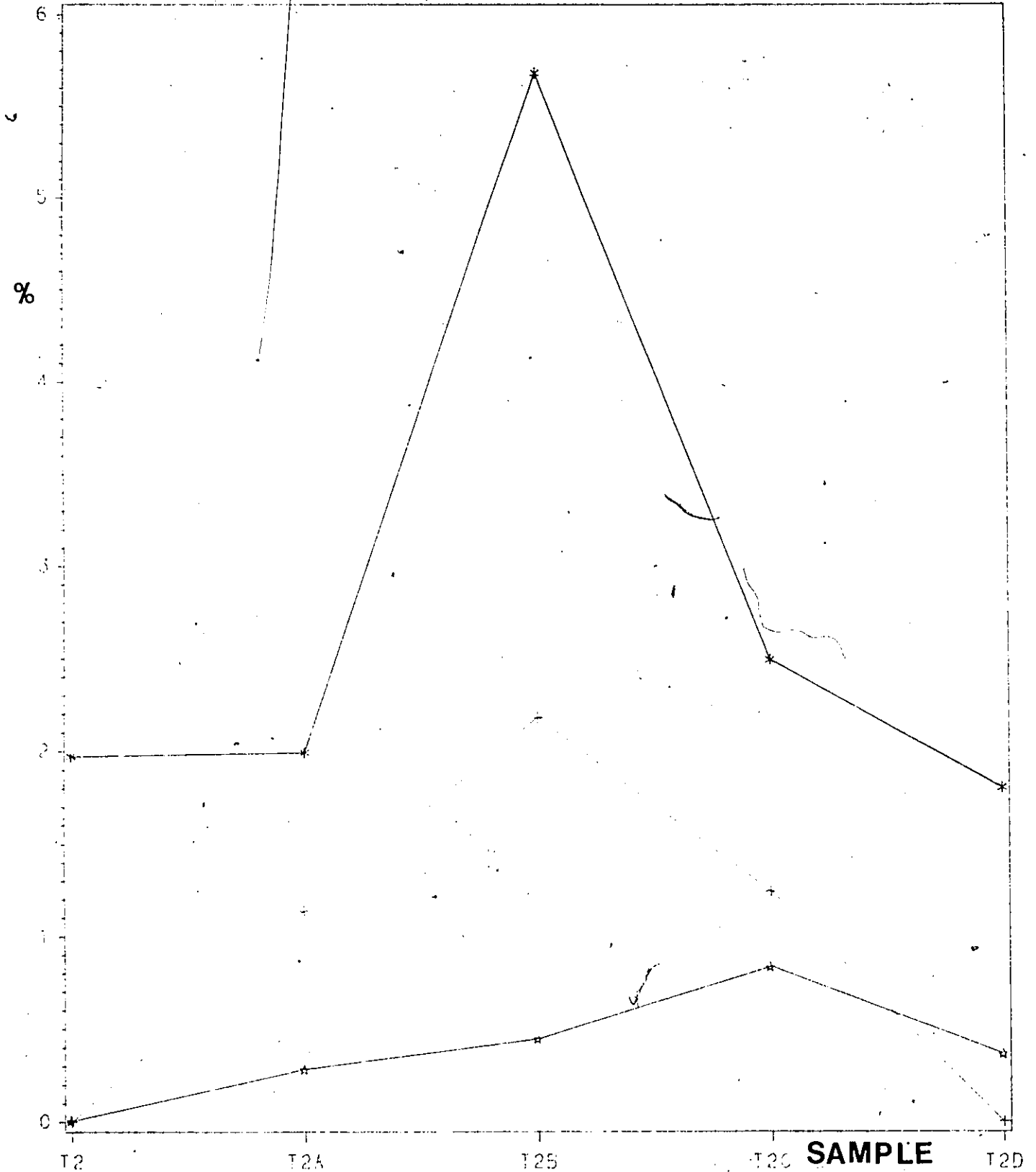


Figure 6.13 Five samples from Tojeira 1 section; which were collected in the same layer, were plotted with respect to their percent values

If frequency distributions of species were binomial distributions, the binomial mean and standard deviation are,

$$\mu = n p$$

and

$$\sigma = (n p q)^{1/2}$$

where,  $\mu$  is a binomial mean,  $\sigma$  is a binomial standard deviation,  $n$  is a grade average of samples and  $p$  is a percent of a microfossil content of a given species. From  $p$  it follows that  $q = 1 - p$ .

$p = 0.0277$ ,  $q = 0.9733$  and  $n = 443$  for Eoguttulina sp.

The mean is

$$\mu = 0.0277 * 443 = 12.27 \text{ -and}$$

the standard deviation is

$$\sigma = (443 * 0.0277 * 0.9733)^{1/2} = 3.45\%$$

$$\text{Binomial noise, } (\sigma_B) = \frac{100 * \sigma}{n}$$

for Eoguttulina sp. is  $\sigma_B = 0.78\%$

$$\text{Noise of variance } (\sigma_N^2) = (1 - c) s^2$$

where  $c$  is constant value which was mentioned before,  $s^2$  is variance of species.

$$\sigma_N = 4.824$$

$$\text{Noise of standard deviation } (\sigma_N) = (\sigma_N^2)^{1/2}$$

$$\sigma_N = 2.19\% \text{ for } \underline{\text{Eoguttulina}} \text{ sp.}$$

The ratio of noise of standard deviation and binomial noise is

$$\sigma_B / \sigma_N = 0.36$$

The same procedures have been followed for all studied series and

the results are shown in Table 6.15. Here, the ratio of the Eoguttulina sp. is larger than 1.00 but, this must be less than 1.00. This discrepancy suggests that the c value of this species was underestimated.

TABLE 6.15 BINOMIAL DISTRIBUTION AND STANDARD DEVIATION OF NOISE

Species	P	Q	C	S <sup>2</sup>	$\bar{n}$	$\frac{\sigma}{\bar{n}}$	$\frac{\sigma}{\bar{n}}$	$\frac{\sigma}{\bar{n}}$	Section	Sample No:
A2: <u>Eoguttulina</u> spp.	0,0277	0,9723	0,76	0,0020		12,27	3,45	0,78	Tojeira.1	30
A3: <u>E. mosquensis</u>	0,2247	0,7753	0,76	0,0160	443	99,54	8,78	1,98		
A10: <u>O. strumosum</u>	0,01929	0,8071	0,50	0,0006		8,54	2,62	0,59		
B1: <u>Eoguttulina</u> spp.	0,0285	0,9715	0,58	0,0002		7,12	2,63	1,05		
B2: <u>E. mosquensis</u>	0,1384	0,8616	0,88	0,0134		34,65	5,46	2,19		
B11: <u>S. tenuissima</u>	0,2575	0,7425	0,90	0,0301	250	64,38	6,91	2,76	Tojeira 2	31
B13: <u>O. strumosum</u>	0,1125	0,8875	0,91	0,0087		28,12	4,99	2,00		
B14: Agglutinants	0,1042	0,8958	0,58	0,0024		26,04	4,83	1,93		
A2: <u>Eoguttulina</u> spp.	0,0220	0,9780	0,48	0,0016		8,99	2,96	0,71		
A3: <u>E. mosquensis</u>	0,2376	0,7624	0,52	0,0147	408	96,92	8,60	2,11	Tojeira 1	41
A10: <u>O. strumosum</u>	0,0239	0,9761	0,60	0,0008		9,76	3,09	0,76		

CHAPTER 7: Correlation between taxa of Tojeira 1 and 2 sections  
and Stam's data

7.1 Introduction

Before correlation, Stam's ( 1986 ) data set ( Table 1.1 and 1.2 ) was transformed with the logit transformation which is close to the probit transformation.

Instead of using Table XI, for logits in Fisher and Yates( 1963 ), a small FORTRAN program was prepared and used as shown below.

The logit transformation,  $Z = ( 1/2 ) \text{LOG}_e ( P/Q )$ ., is equivalent to the R, Z transformation with  $R = 2P - 1$ . For values of  $P < 0.5$  logits are negative and numerically equal to the tabulated values for  $1 - P$ .

```

DIMENSION P(40,40),Q(40,40),R(40,40),A(40,40),Z(40,40)
READ (5,1000) N,M
DO 200 I = 1,N
  READ ( 5,1001) ( P ( I,J),J = 1,M)
200 CONTINUE
1000 FORMAT ( 2I2)
1001 FORMAT ( 14F4.1)
DO 100 I=1,N
  DO 101 J=1,M
    Q(I,J)= 100.0-P(I,J)
  
```

TABLE 7.1 CORRELATION COEFFICIENTS OF 14 SPECIES IN TOJEIRA 1 SECTION (STAM'S DATA)

	A1	A2	A3	A4	A5	A6	A7	A8	A9	A10	A11	A12	A13	A14
A1														
A2	-0.586 0.001													
A3	0.394 0.028	-0.679 0.000												
A4	0.182 0.328	-0.450 0.011	0.528 0.002											
A5	0.110 0.555	-0.359 0.047	0.433 0.015	-0.010 0.957										
A6	-0.223 0.228	0.280 0.127	-0.438 0.014	0.029 0.876	-0.380 0.035									
A7	0.147 0.429	0.004 0.982	-0.175 0.348	-0.182 0.326	-0.269 0.143	0.461 0.009								
A8	-0.165 0.376	0.225 0.224	-0.395 0.028	-0.033 0.861	-0.482 0.001	0.192 0.300	0.065 0.727							
A9	-0.170 0.359	0.197 0.289	-0.193 0.299	-0.067 0.718	-0.239 0.196	-0.181 0.329	-0.060 0.751	0.181 0.331						
A10	0.393 0.029	-0.494 0.005	0.462 0.009	0.024 0.898	0.341 0.061	-0.099 0.595	0.067 0.720	0.055 0.767	-0.280 0.127					
A11	0.436 0.014	-0.659 0.000	0.591 0.001	0.303 0.097	0.221 0.232	-0.246 0.182	0.019 0.921	-0.222 0.231	-0.258 0.161	0.402 0.025				
A12	0.081 0.666	-0.149 0.423	0.407 0.023	0.024 0.899	0.158 0.396	-0.323 0.076	-0.103 0.582	-0.118 0.529	0.379 0.036	0.212 0.252	0.689 0.714			
A13	-0.660 0.000	0.731 0.000	-0.625 0.000	-0.423 0.018	-0.206 0.267	0.216 0.243	-0.101 0.590	0.101 0.587	0.027 0.886	-0.484 0.006	-0.688 0.000	-0.199 0.283		
A14	0.013 0.946	0.300 0.101	-0.451 0.011	-0.429 0.016	-0.148 0.427	-0.227 0.219	0.029 0.879	0.043 0.819	0.166 0.372	-0.368 0.042	-0.318 0.082	-0.074 0.692	0.188 0.312	

```

R(I,J)= P(I,J)/Q(I,J) .
A(I,J)= LOG(R(I,J))
Z(I,J)= A(I,J)/2.0
101 CONTINUE
100 CONTINUE
DO 90 I =1,N
WRITE (6,2000) (Z(I,J), J =1,M )
90 CONTINUE
2000 FORMAT ( 14 ( 1x,F7.4))
STOP
END

```

NOTE : Table 1.1 and 1.2 was thought of as a matrix ( N = 31, M = 14 )

Logit transformed data were used as input to the CORR SAS package program to calculate correlation coefficients.

## 7.2 Correlation of taxa in Tojeira 1 section

Significant positive and negative correlation coefficients between the various taxa are given in Table. 7.1.

Pseudolmarckina rjasanensis, Epistomina mosquensis and Epistomina uhligi are known to characterize deeper shelf deposits ( Pazdro, 1969; Gradstein, 1983; Lloyd, 1970). All species of this group are positively correlated to one the other.

Spirillina tenuissima fits in well and actually confirms the idea of this

species preferring shallow water ( Stam, 1986 ). Spirillina infima is positively correlated with Spirillina tenuissima but Spirillina elongata is negatively correlated to it. This was not really expected. However, this correlation coefficient is very close to 0 and it is not significantly different from zero.

Lenticulina muensteri muensteri is positively correlated with the Lenticulina sp. and negatively correlated with Epistomina sp. Epistomina sp. has a moderate degree of positive correlation with Epistomina mosquensis and a negative correlation with Lenticulina muensteri and Nodosaria/Dentalina sp. which indicates that they are closer to the deep water association rather than the shallow water one.

There is no correlation between Lenticulina muensteri and Epistomina uhligi, or between Epistomina uhligi and Spirillina infima.

The group formed by Epistomina sp. and Spirillina elongata is, considering their negative correlations with the shallow water group, closer to the deeper group.

The correlation coefficients of the species are usually low, but, as seen in Table 7.1, Spirillina infima has a high positive correlation with Eoguttulina sp. Epistomina mosquensis with agglutinants and Epistomina mosquensis with Eoguttulina have a high negative correlation.

### 7.3 Correlation of taxa in Tojeira 2 section

Pseudolamearckina rjasanensis, Epistomina mosquensis, Ophtalmidium strumosum and agglutinants have a high positive correlation (Table 7.2)

TABLE 7.2 CORRELATION COEFFICIENTS OF 14 SPECIES IN TOJEIRA 2 SECTION (STAM'S DATA)

	B1	B2	B3	B4	B5	B6	B7	B8	B9	B10	B11	B12	B13	B14
B1														
B2	-0.483 0.007													
B3	0.085 0.653	0.316 0.089												
B4	0.366 0.047	-0.424 0.020	0.444 0.014											
B5	-0.161 0.396	-0.010 0.960	-0.167 0.377	0.120 0.527										
B6	0.040 0.834	-0.236 0.210	-0.149 0.432	0.127 0.503	0.045 0.813									
B7	-0.066 0.728	-0.015 0.936	0.265 0.157	0.110 0.562	-0.178 0.348	0.117 0.539								
B8	-0.445 0.014	0.788 0.000	0.268 0.153	-0.367 0.046	0.030 0.875	-0.152 0.423	0.055 0.771							
B9	-0.034 0.859	0.096 0.613	-0.325 0.080	-0.241 0.200	-0.023 0.902	-0.064 0.736	-0.141 0.457	0.001 0.996						
B10	0.455 0.011	-0.662 0.000	-0.010 0.960	0.325 0.080	-0.188 0.319	0.019 0.921	-0.145 0.445	-0.673 0.000	0.061 0.750					
B11	0.502 0.005	-0.909 0.000	-0.369 0.045	0.343 0.064	-0.057 0.765	0.188 0.320	-0.131 0.491	-0.827 0.000	-0.021 0.911	0.749 0.000				
B12	0.454 0.012	-0.257 0.170	-0.213 0.259	0.036 0.851	0.070 0.714	-0.263 0.160	-0.309 0.097	-0.172 0.363	-0.325 0.080	0.046 0.807	0.264 0.158			
B13	-0.385 0.036	0.870 0.000	0.163 0.390	-0.515 0.004	0.138 0.468	-0.236 0.210	0.031 0.869	0.774 0.000	0.041 0.829	-0.726 0.000	-0.899 0.000	-0.018 0.923		
B14	-0.460 0.011	0.699 0.000	0.138 0.467	-0.533 0.002	0.153 0.418	-0.106 0.578	0.048 0.801	0.616 0.000	-0.100 0.601	-0.632 0.000	-0.754 0.000	-0.120 0.528	0.745 0.000	

Consistent with this are the negative correlations between Epistomina mosquensis, Eoguttulina sp. and Spirillina infima.

The group formed by Epistomina mosquensis, Pseudolamerckina rjasanensis, Ophthalmidium strumosum and agglutinants has a preference for relatively deeper water. A second group formed by Eoguttulina sp., Spirillina infima and Spirillina tenuissima has a preference for shallow water. Epistomina mosquensis is negatively correlated with Lenticulina and Epistomina sp.

Epistomina uhligi, Lenticulina muensteri, Nodosaria/Dentalina sp., Spirillina elongata have a preference for deeper water. Lenticulina, Epistomina sp. and Ophthalmidium carinatum have a preference for shallow water as suggested by the correlations in Table 7.2.

The result of correlating taxa within the Tojeira 1 and Tojeira 2 sections shows that the benthonic foraminifers microfauna is the same for both sections, but species of Tojeira 2 are more strongly correlated than the species of Tojeira 1 as seen in Tables 7.1 and 7.2.

In general, the correlations of this section, based on logits of proportion values for the species, confirm the correlations previously obtained by Stam ( 1986 ) who used Drooger's ( 1982 ) BALANC and DISTUR computer programs.

## CHAPTER 8: Conclusions

In this new geostatistical approach to biostratigraphic sampling and spatial analysis, two stratigraphic sections were studied in detail. They are the so - called Tojeira 1 and Tojeira 2 sections located in central Portugal. The geology and macropaleontology of area have been studied since 1880. The lithology and micropaleontology have been clearly established by Stam ( 1986 ).

For statistical analysis, the count is considered to be the numerical proportion of a taxon from an assemblage of foraminifera. It is characterized as a series. These series have a logical order ( from bottom to top or from top to bottom of sections ) and are related to time. Time series analysis shows how a time series variable (  $X_t$  ) can be related to its own past values or that there may be an interdependence in the sense that some taxon property ( its proportion ) in a certain sample is strongly related to the one(s) in the sample(s) immediately below, even if the proportions of species are zero in some of the samples.

The duration of time from Middle to Late Jurassic age is approximately 11 m.y. Microfossil sample data for this time interval were statistically analyzed as time series. The samples plotted were proportion values of species. At Tojeira 1 section, the proportions of taxon Epistomina mosquensis and Ophthalmisium strumosum seem to show an increasing trend, and Eoguttulin sp. a decreasing trend towards the late Jurassic age (Figs. 8.1 - 8.3) with Stam's data and Figs.8.4

Figure 8.1 Time series of 31 Eoguttulina sp. percent values from Tojeira 1 section. Data are autocorrelated according to Table 1.1

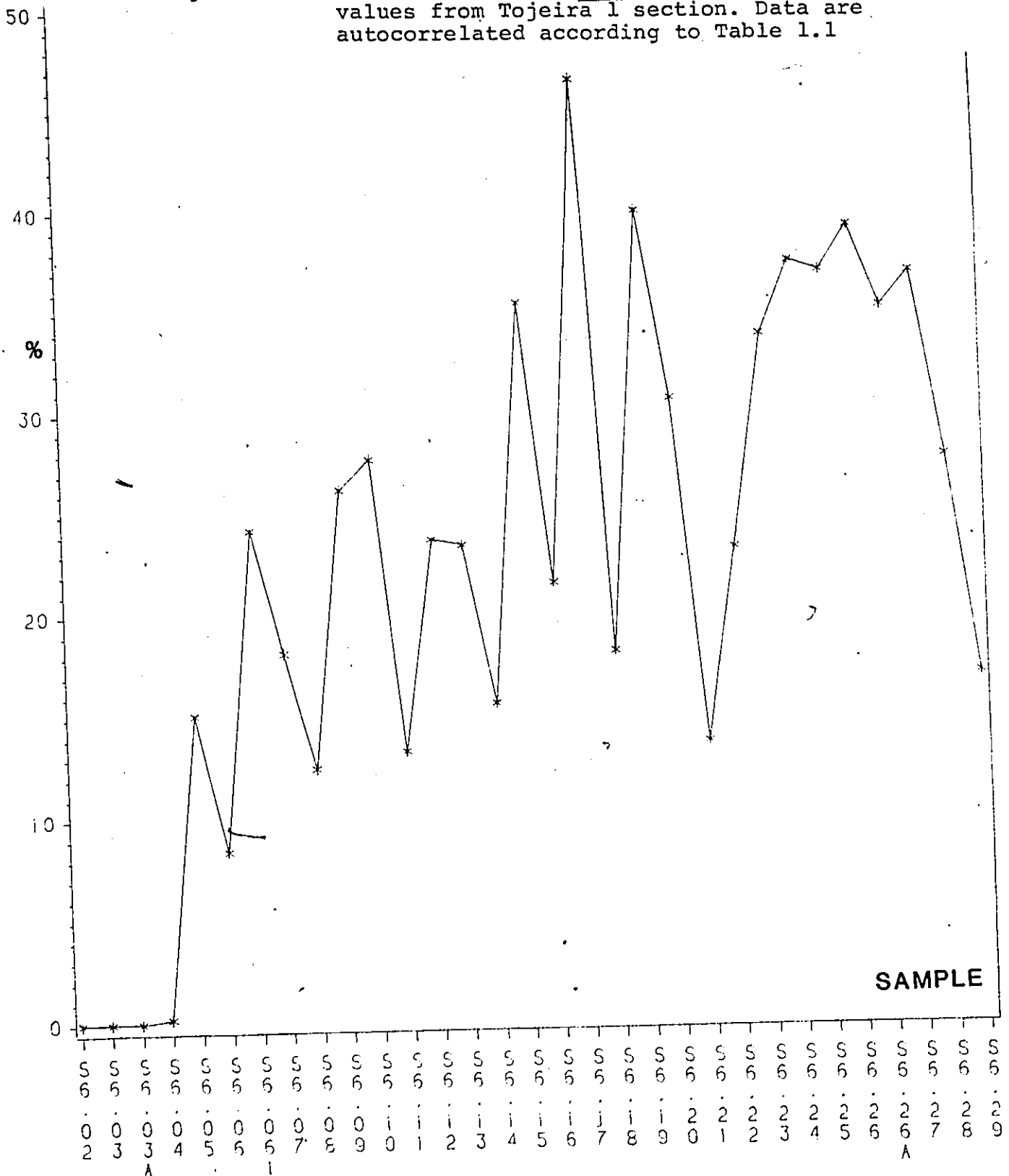


Figure 8.2 Time series of 31 E. mosquensis percent values from Tojeira 1 section. Data are autocorrelated according to Table 1.1

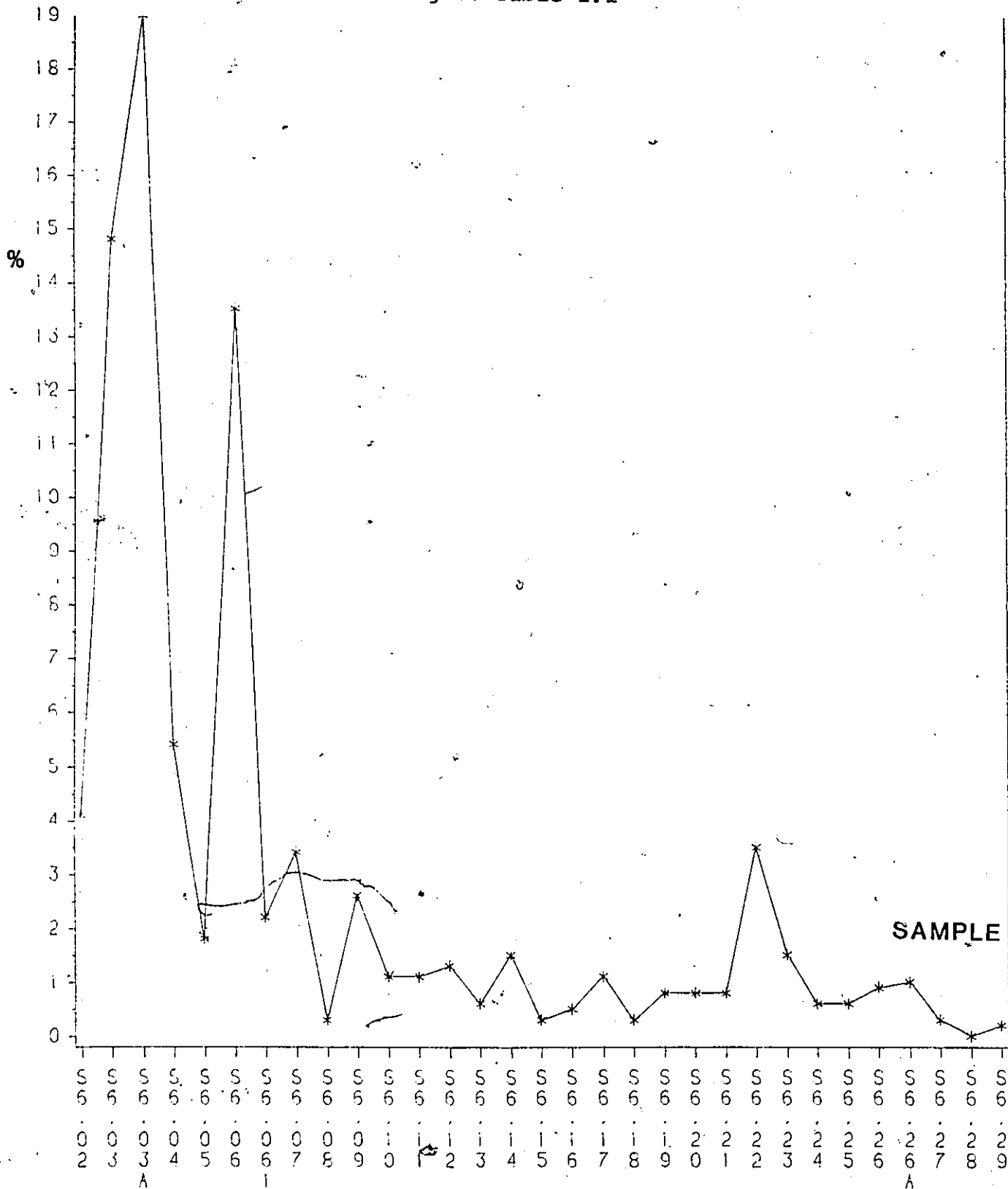
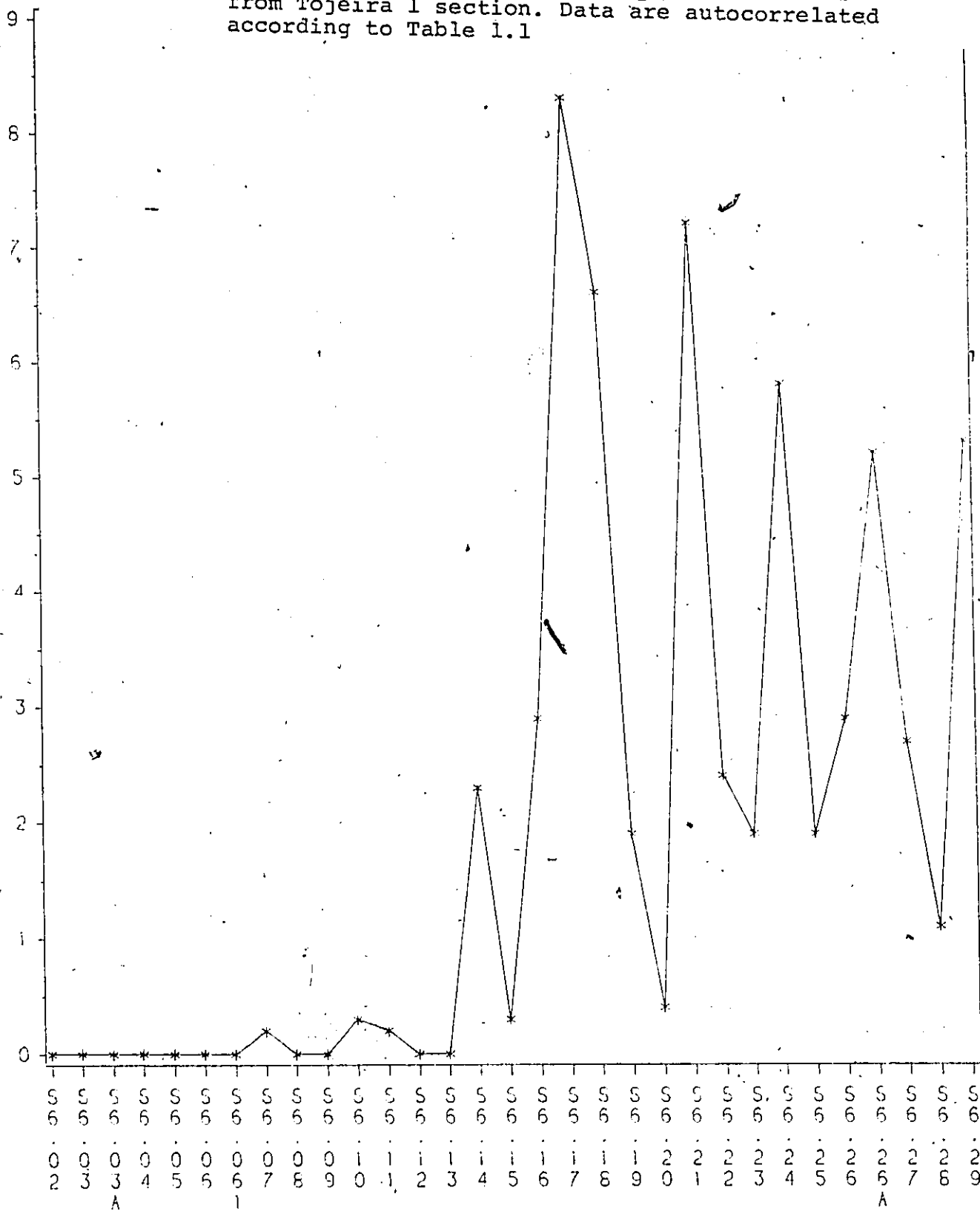


Figure 8.3 Time series of 31 O. strumosum percent values from Tojeira 1 section. Data are autocorrelated according to Table 1.1



- 8.6 with enlarged data ). At Tojeira 2 section, the proportions of taxa Epistomina mosquensis, Ophthalmidium strumosum and agglutinants may show an increasing trend and Eoguttulina sp. and Spirillina tenuissima show a decreasing trend from Middle Jurassic to late Jurassic age ( Figs. 8.4 - 8.8)

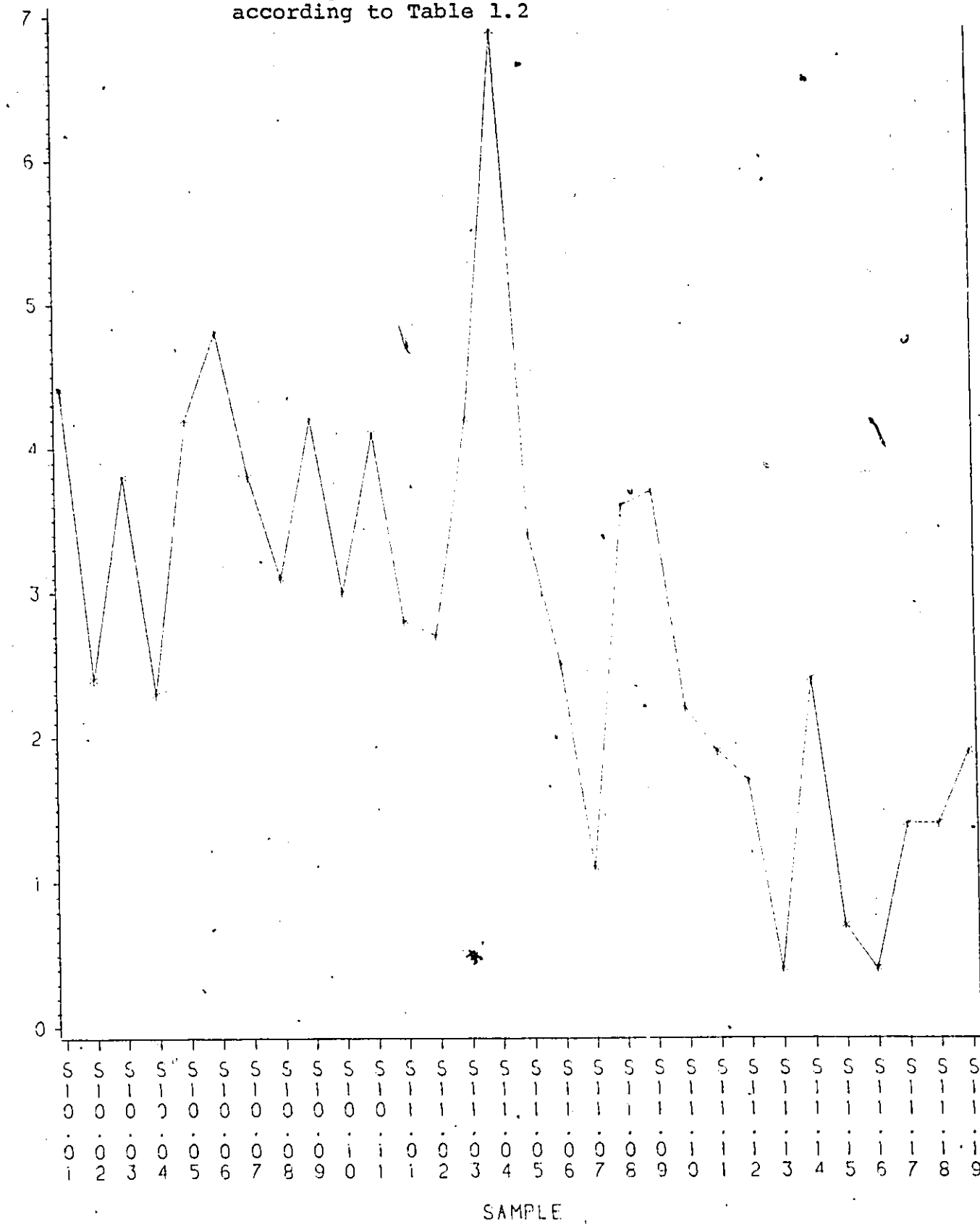
The ARIMA process, formed from a combination of autoregressive and moving average processes, was used to calculate autocorrelation coefficients for 14 species in both the Tojeira 1 and Tojeira 2 sections ( Appendix I ).

Before starting the new geostatistical model, the data was tested in terms of a time series model. The three time series models that were considered were the AR ( p, q ), p = 1 and q = 0, autoregressive model, the MA ( p, q ), p = 0 and q = 1, moving average model and the ARMA ( p, q ) p = 1, 2 and q = 1, 2 autoregressive and moving average model. It was determined that only the AR ( p, q ) and ARMA ( p, q ) models explained the data suitably. This was because these two models are signal - plus - noise models.

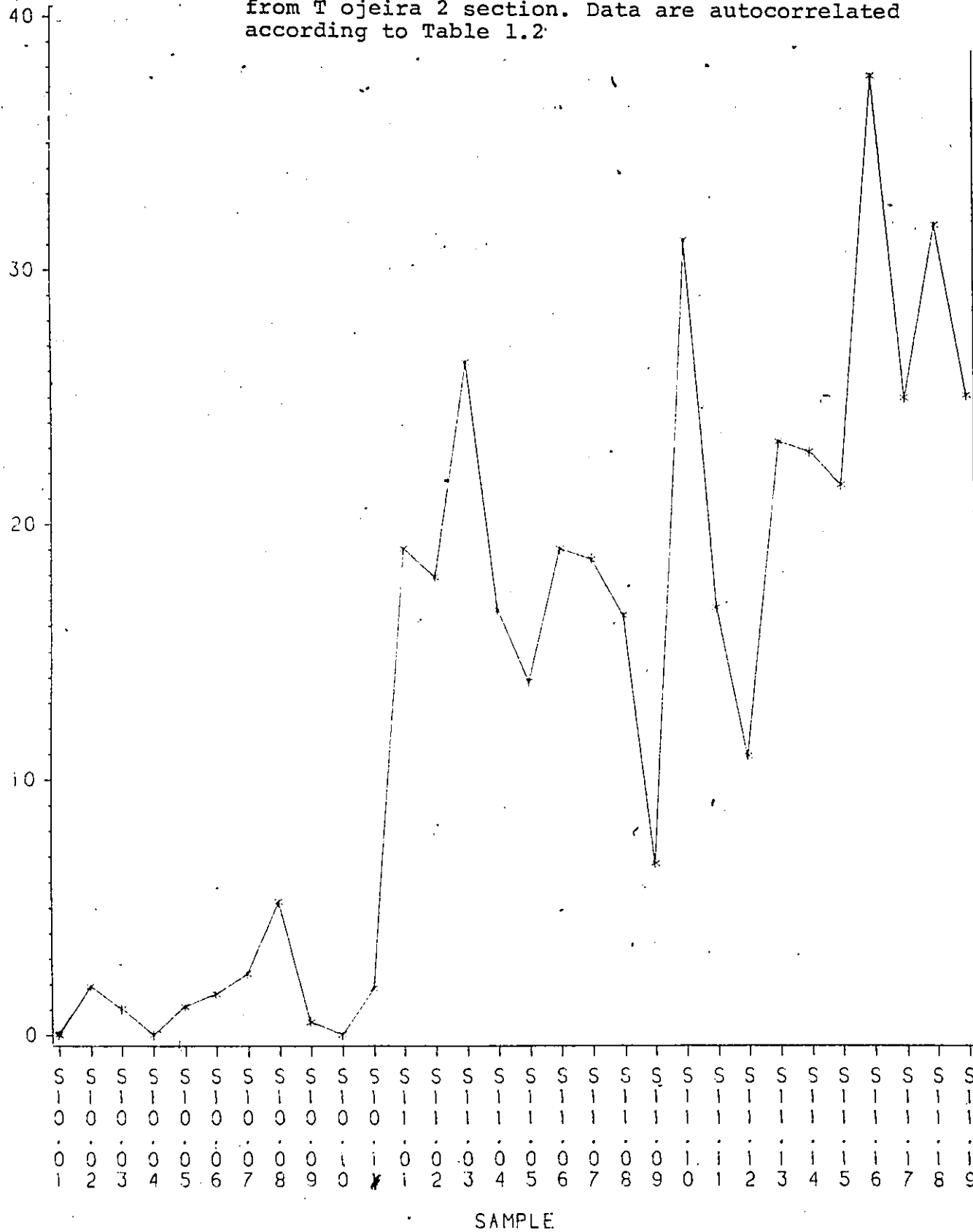
The preparation and counting procedures play a primary role in the magnitude of the white noise ( error ). The white noise was tested according to t and chi - squared tests, as was the periodogram which was calculated by the SAS SPECTRA procedure.

If the observations during the count really are random, the statistical error may be considered to be equal to the theoretical standard deviation of the binomial distribution model. The ratio of the theoretical standard deviation from the binomial distribution to the

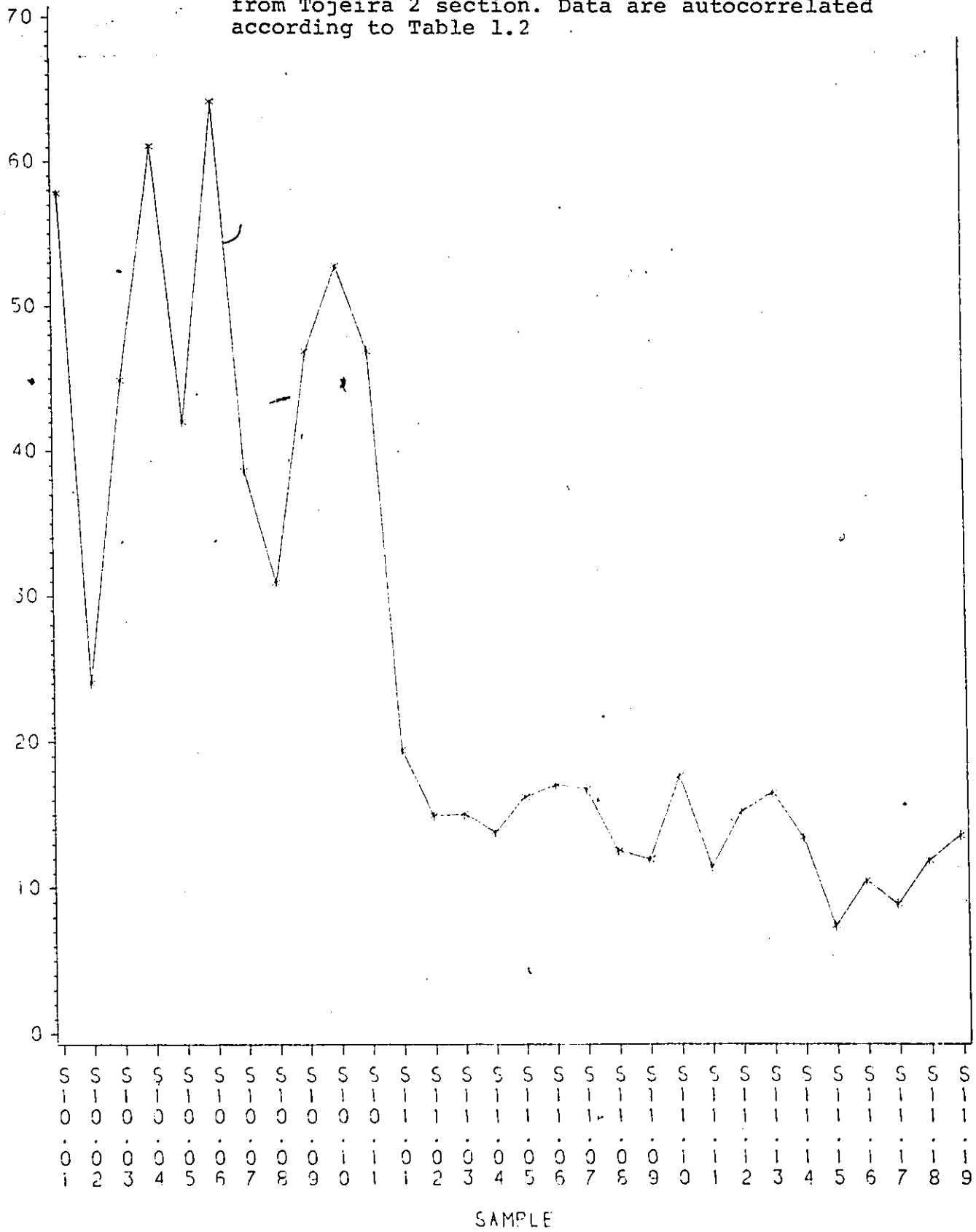
Figure 8.4 Time series of 30 Eoguttulina sp. percent values from Tojeira 2 section. Data are autocorrelated according to Table 1.2



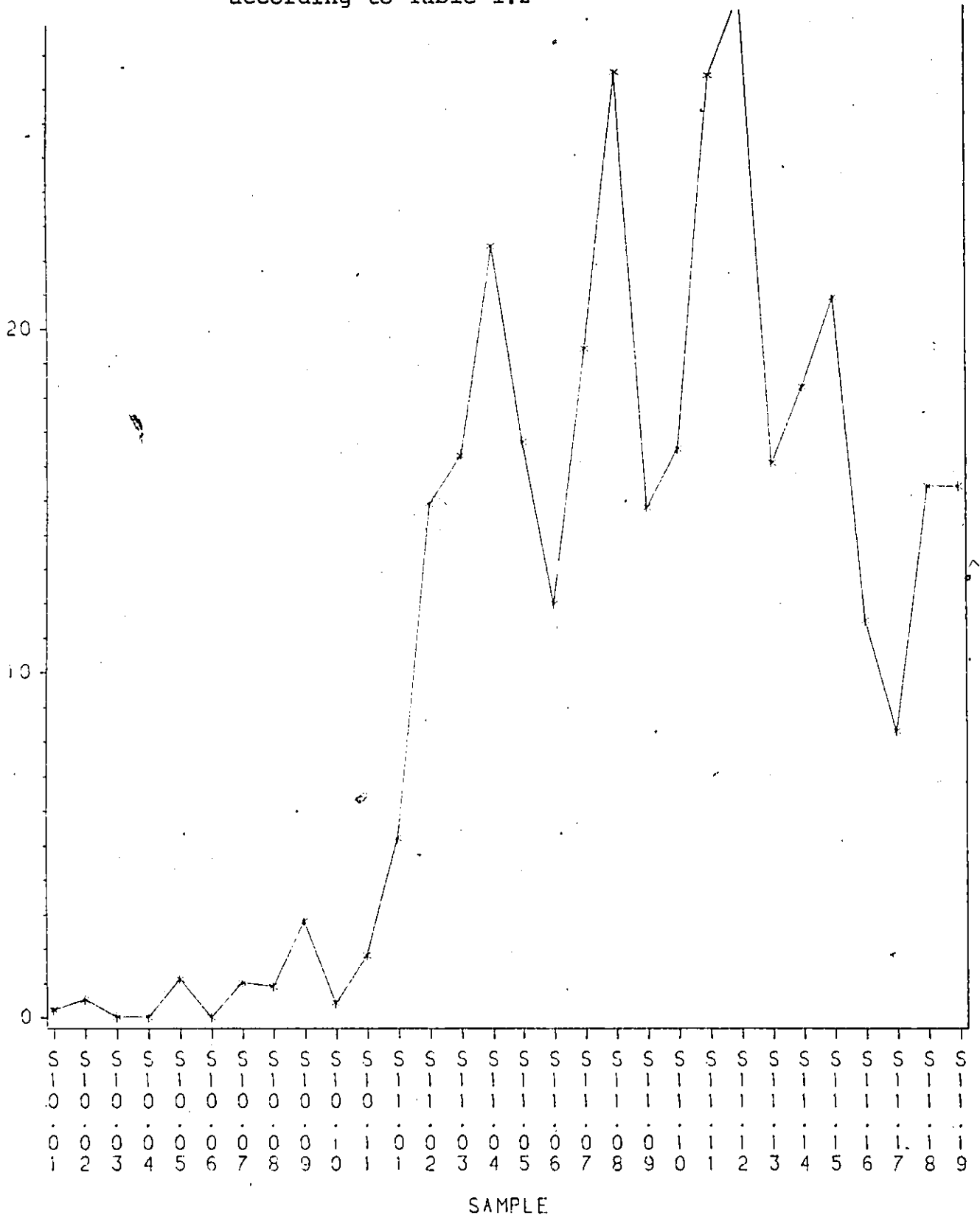
B2 Figure 8.5 Time series of 30 E: mosquensis percent values from T ojeira 2 section. Data are autocorrelated according to Table 1.2



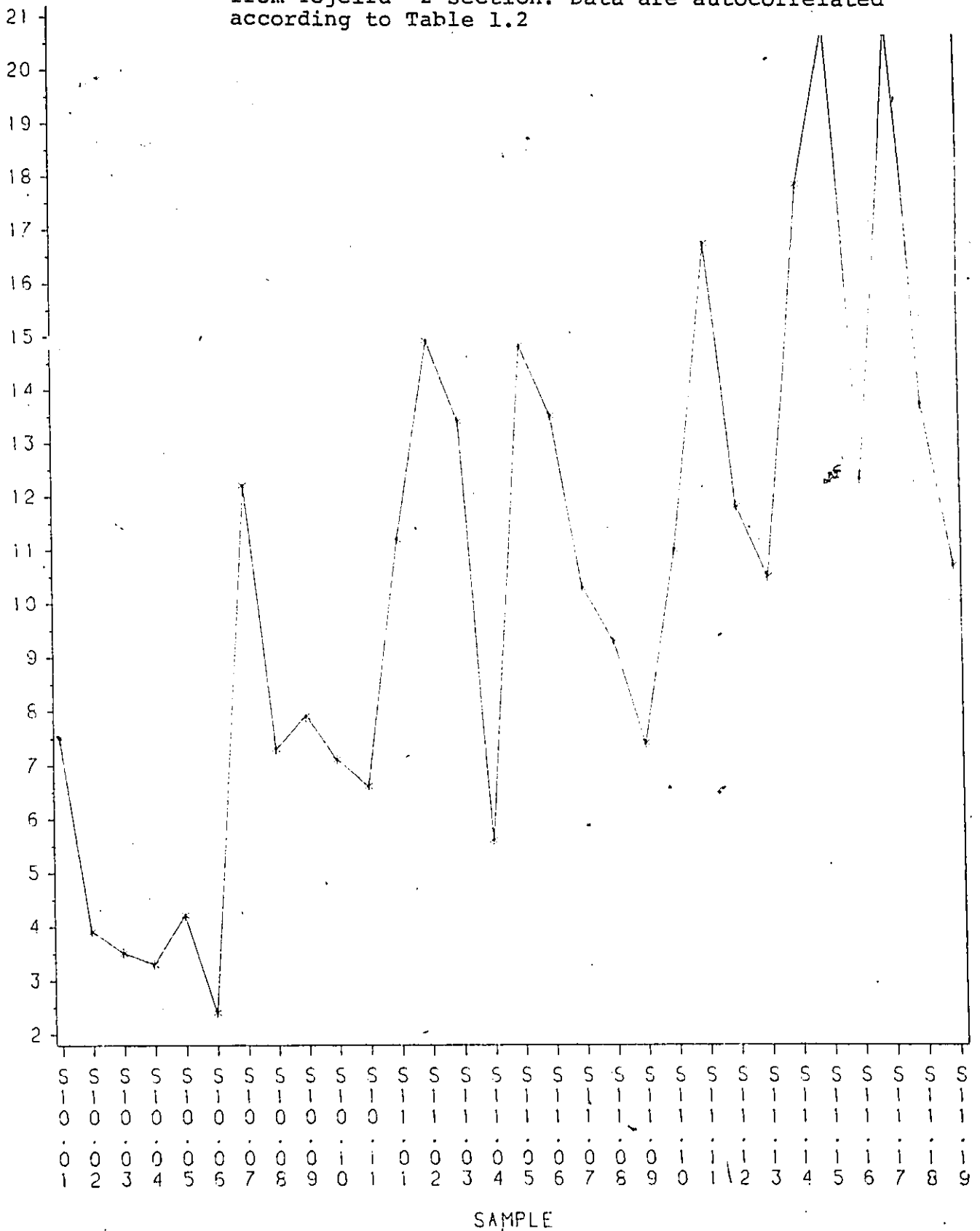
B11 Figure 8.6 Time series of 30 *S. tenuissima* percent values from Tojeira 2 section. Data are autocorrelated according to Table 1.2



B13 Figure 8.7 Time series of 30 O. strumosum percent values from Tojeira 2 section. Data are autocorrelated according to Table 1.2.



B14 Figure 8.8 Time series of 30 Agglutinants percent values from Tojeira 2 section. Data are autocorrelated according to Table 1.2



standard deviation of the probnormal distribution ( geostatistical model) was calculated and shown in Table 6.15.

A good starting point for this type of modeling is to construct the covariance of a pattern. The covariance can be converted into an autocorrelation coefficient by using standard statistical formulae or the SAS package program ARIMA process.

The autocorrelation function ( $r_l$ ) is exponential for all directions. The autocorrelation coefficients ( $r_l$ ) plotted along the vertical axis with a logarithmic scale. The geostatistical model or the signal - plus noise model with  $r = c \exp.(-al)$  where  $a$  and  $c$  are two constants characterizing the stationary series. The  $a$  and  $c$  are estimated from a correlogram ( see Figs.4.8, 4.13, 4.18, 4.21, 6.4, 6.5 and 6.6 ). If noise is absent,  $c = 1$  and the line passes through the point  $r(0) = 1$  for zero lag. If the series consists of noise only,  $c = 0$ .

Estimated  $a$  and  $c$  values were used for the calculation of the  $F_L$  function ( Eq.54). The  $F$  function was also determined experimentally by averaging an increasingly greater number of adjacent values of species. This gave the factors  $F_N$  calculated by the F FORTRAN program ( see Appendix I ) for relative variance of the mean multiplied by the number of observations (  $N$  ). For uncorrelated data,  $F_N$  is equal to one. The  $F_N$  and  $F_L$  functions were plotted against the lag as an observation and as a theoretical curve. They all show a good fit for the species studied. ( see Figs. 4.9, 4.14, 4.19, 4.27, 4.28,4.29, 4.30, 4.31, 6.7, 6.8, and 6.9 ).

In the model used for this study, the frequency distribution of a

Figure 8.9 Time series of 41 Equttulina sp. percent values from Tojeira 1 section. Data are autocorrelated according to Table 6.1

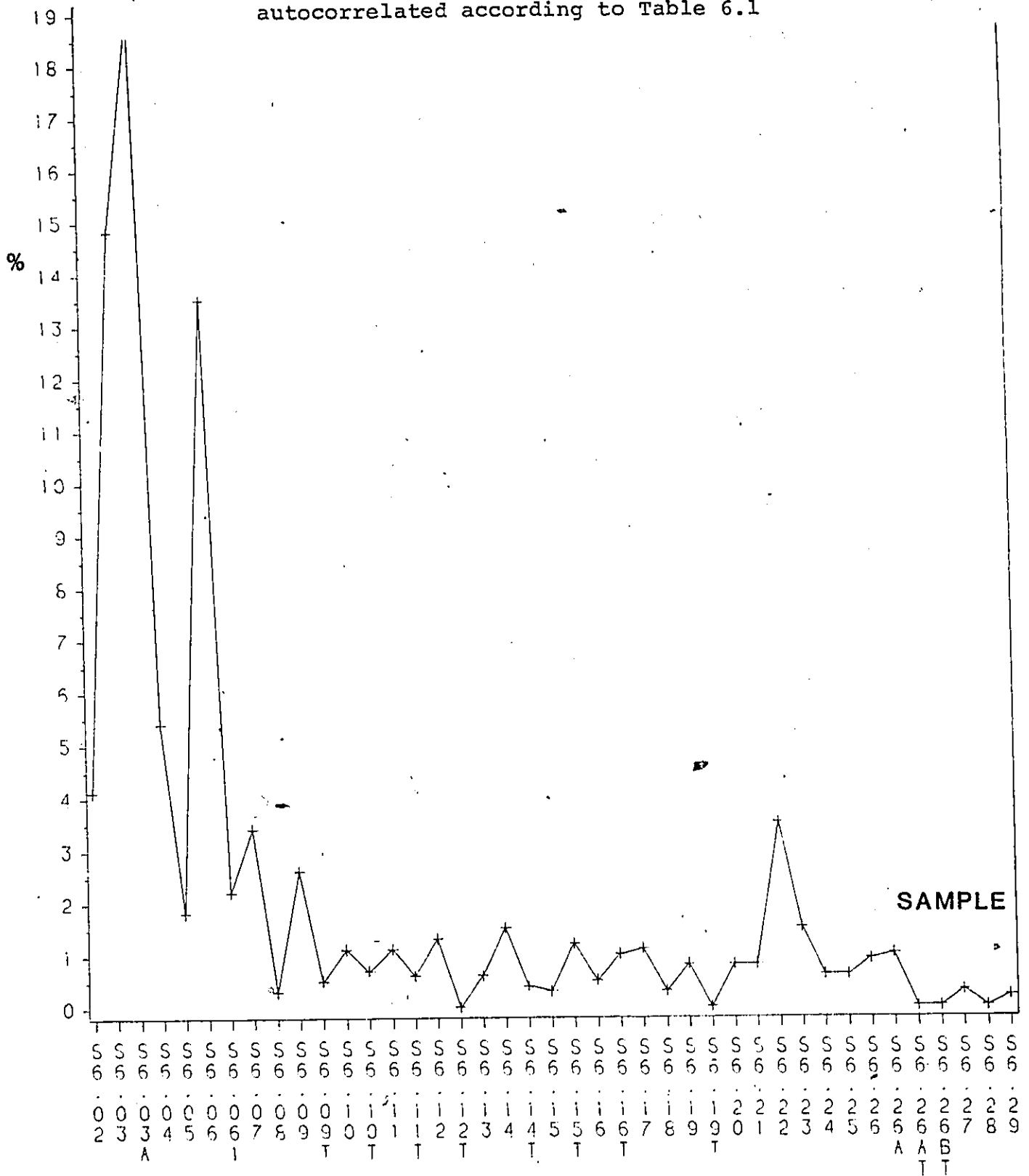


Figure 8.10 Time series of 41 *E. mosquensis* percent values from Tojeira 1 section. Data are autocorrelated according to Table 6.1

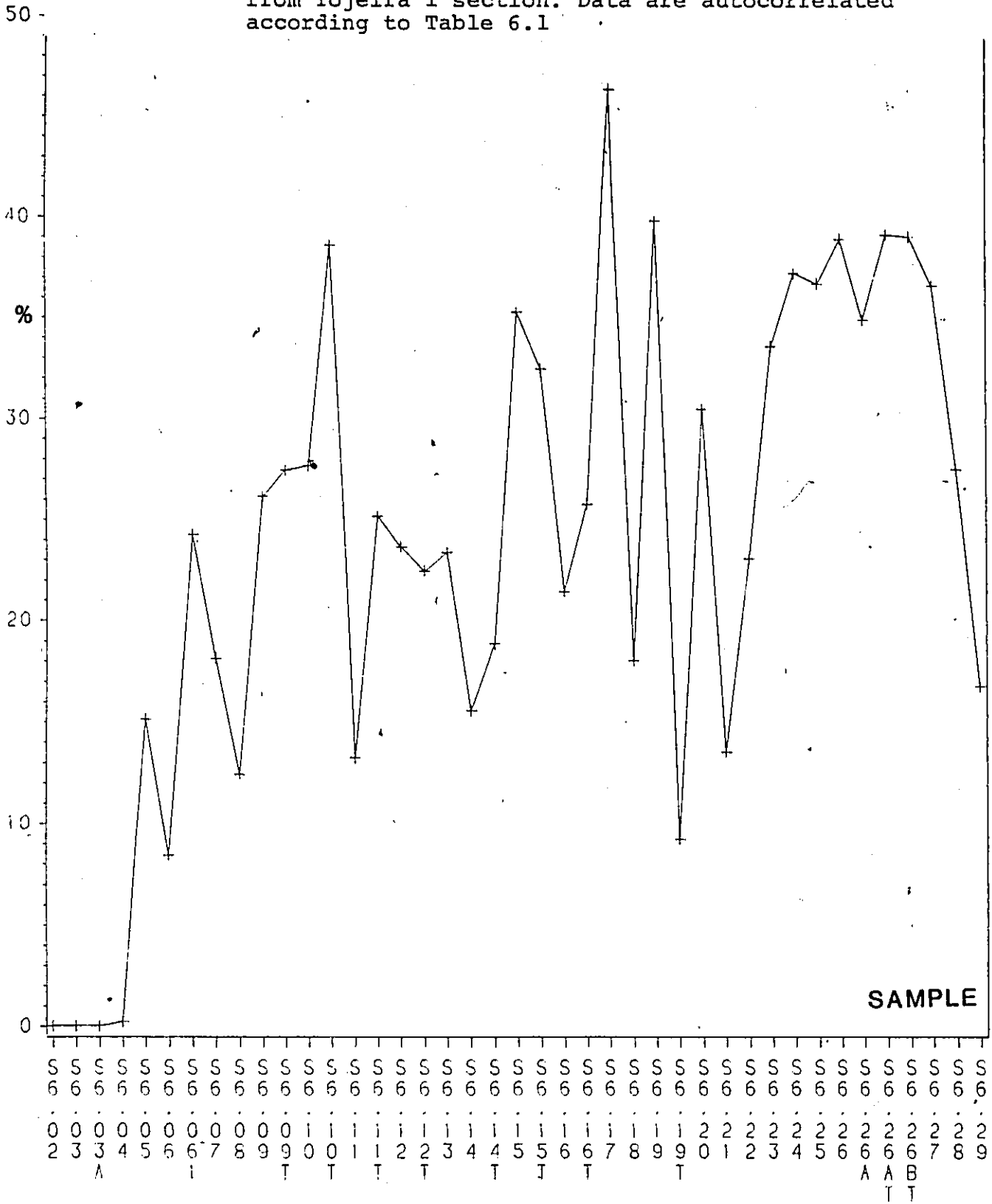
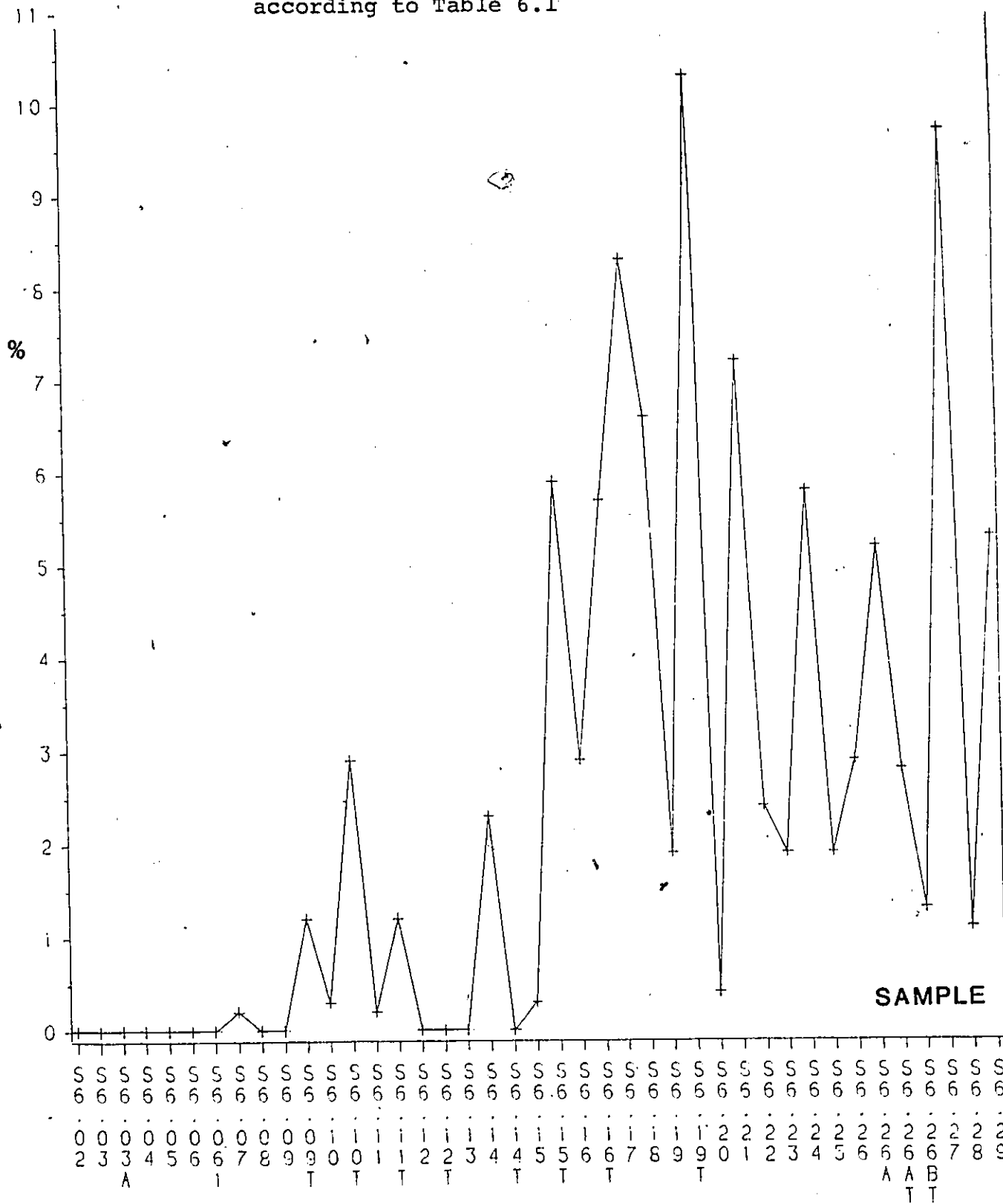


Figure 8.11 Time series of 41 *O. strumosum* percent values from Tojeira 1 section. Data are autocorrelated according to Table 6.1

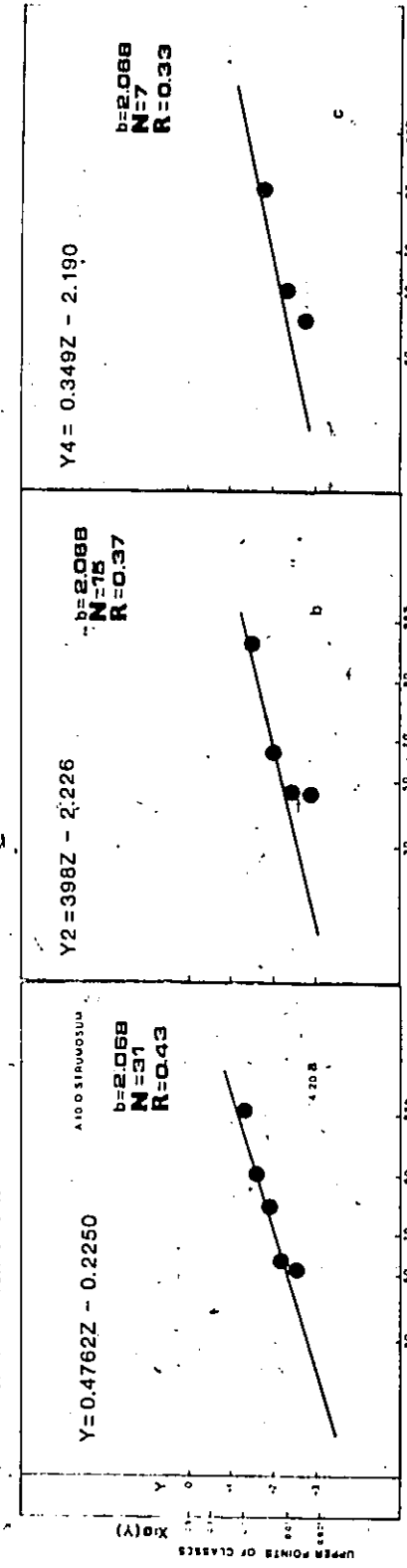
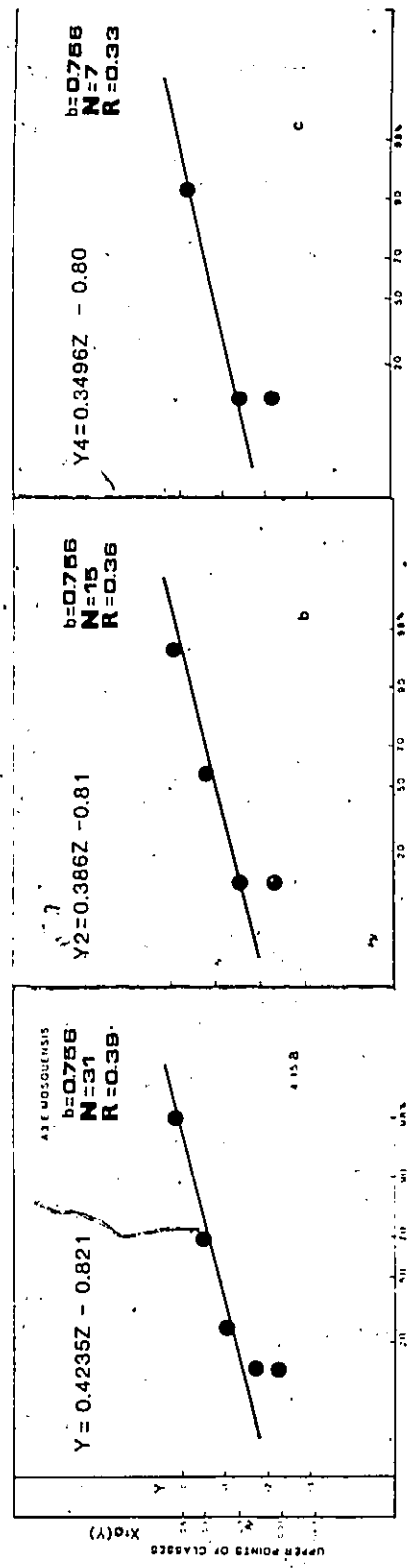
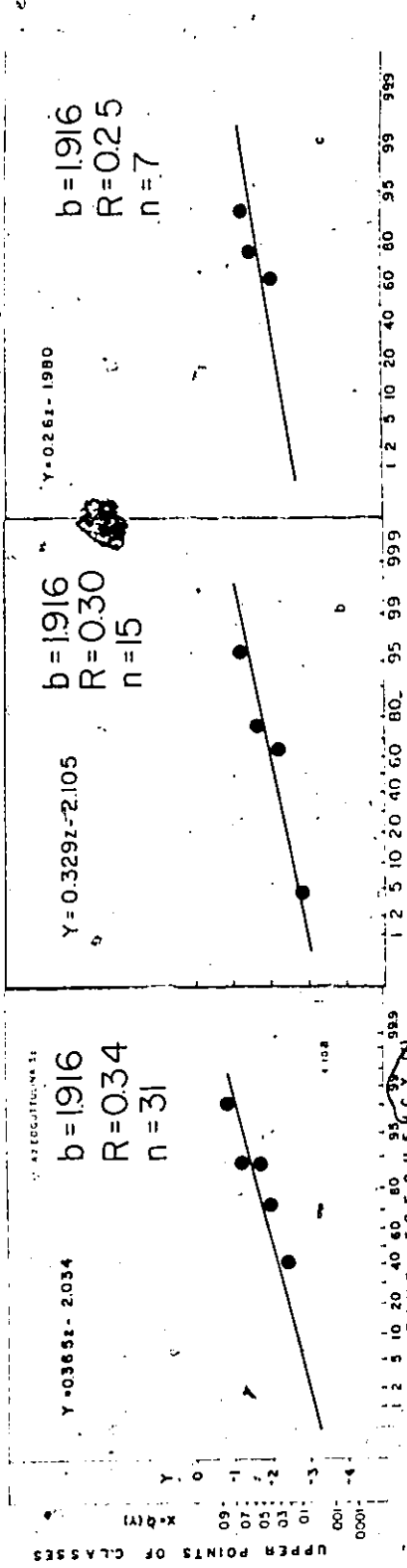


random variable ( $X_t$ ) in Eq. 60 becomes a straight line if the function is plotted on paper that has a normal probability scale both for the ordinate and for the abscissa representing the cumulative frequency percentage. If the model is good, a straight line is also obtained by plotting the probit of the proportion value of the taxon ( see probit transformation tables for each taxon ) on normal probability paper. This procedure facilitates the comparison between the model given by Eq.60 ( the theoretical distribution ) and any observed frequency distribution for a sample obtained by superimposing a grid on the pattern and computing the average proportion values of groups of adjoining original samples.

In our statistical model study three different data sets were applied successively to both sections.

First, Stam's data ( Tables 1.1 and 1.2 ) were used for application of the geostatistical model. Nine examples of the graphical test of the model are shown in Fig. 8.12. These nine plots are for the previously discussed foraminifer species of Upper Jurassic age in the Tojeira 1 section. The first set of 3 plots is for Eoguttulina sp.. The mean  $\bar{x} = 0.0277$  has a corresponding variance of  $S^2 = 0.00201$  for 31 samples,  $S^2_2 = 0.00154$  for 15 samples which are combinations of pairs of adjacent single samples and  $S^2_4 = 0.00052$  for 7 samples which are combinations of groups of four adjacent single samples. These data yield the values of  $b$  and  $R$  shown in Fig. 8.12. The second set of 3 plots is for Epistomina mosquensis with the mean  $\bar{x} = 0.2247$  and variances  $S^2 = 0.01654$ ,  $S^2_2 = 0.01232$  and  $S^2_4 = 0.00947$  for 31,

Figure 8.12 Experimental frequency distributions and their theoretical distributions for 31, 15 and 7 values of Eoguttulina sp., E. mosquensis and O. strumosum. The values of b's and R's obtained from their mean and variance for Tojeira 1 section with Stam's data



15 and 7 samples, respectively. The third set of 3 plots are for Ophthalmidium stromosum with mean  $\bar{x} = 0.01929$ , and variances  $S^2 = 0.02445$ ,  $S^2_2 = 0.00042$  and  $S^2_4 = 0.00032$ .

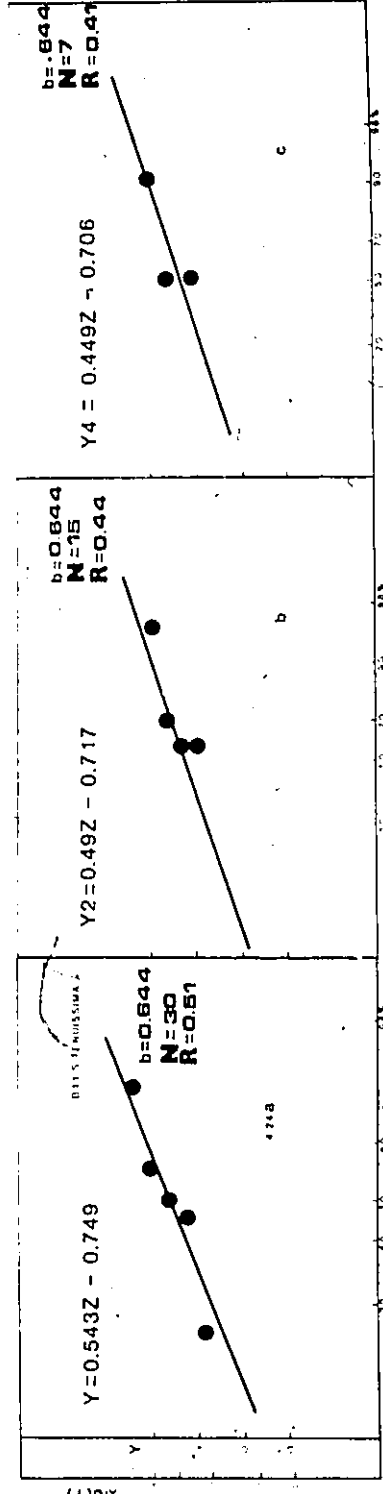
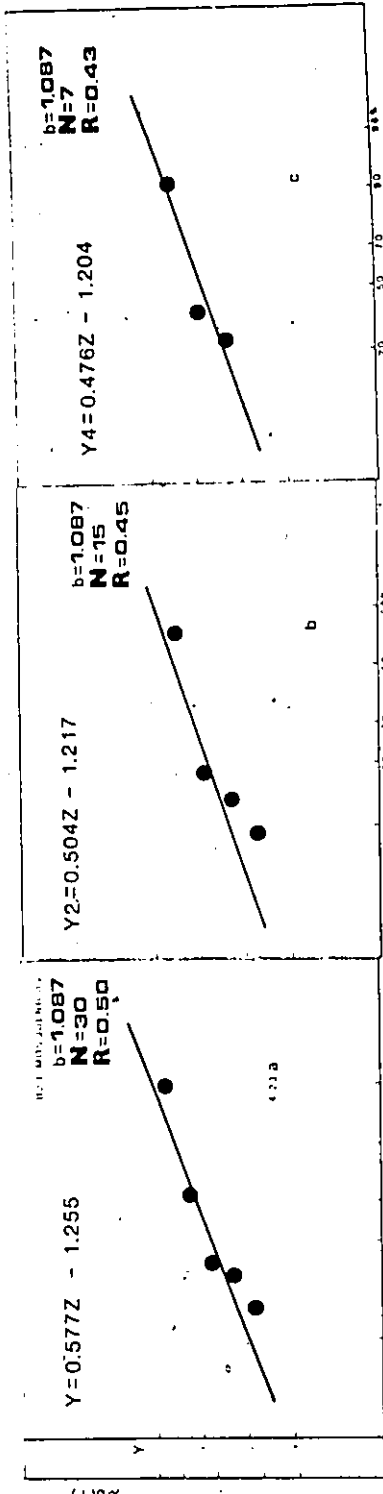
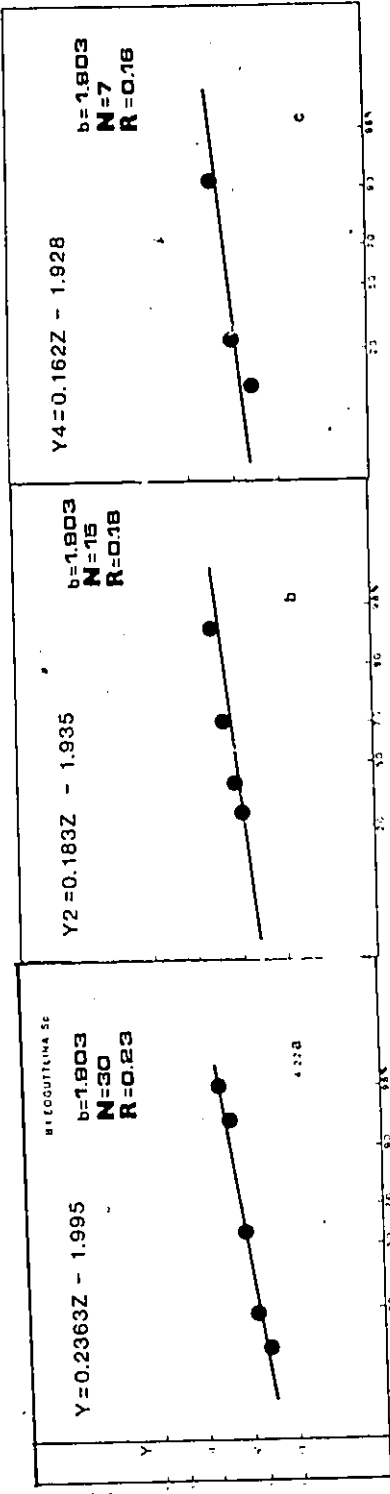
The next fifteen examples of the model are shown in Fig. 8.13. These fifteen plots of species of Middle - Upper Jurassic-age are for the Tojeira 2 section.

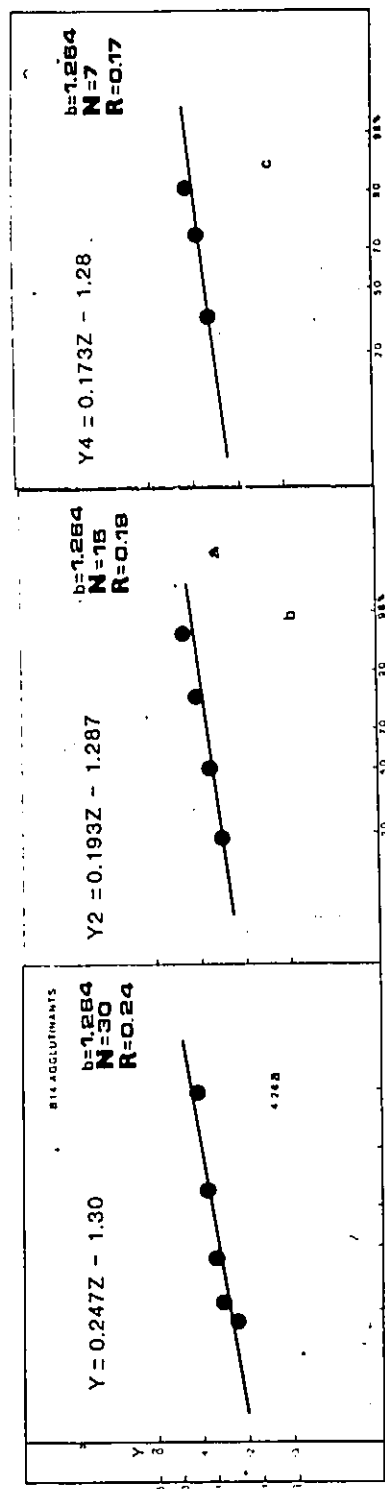
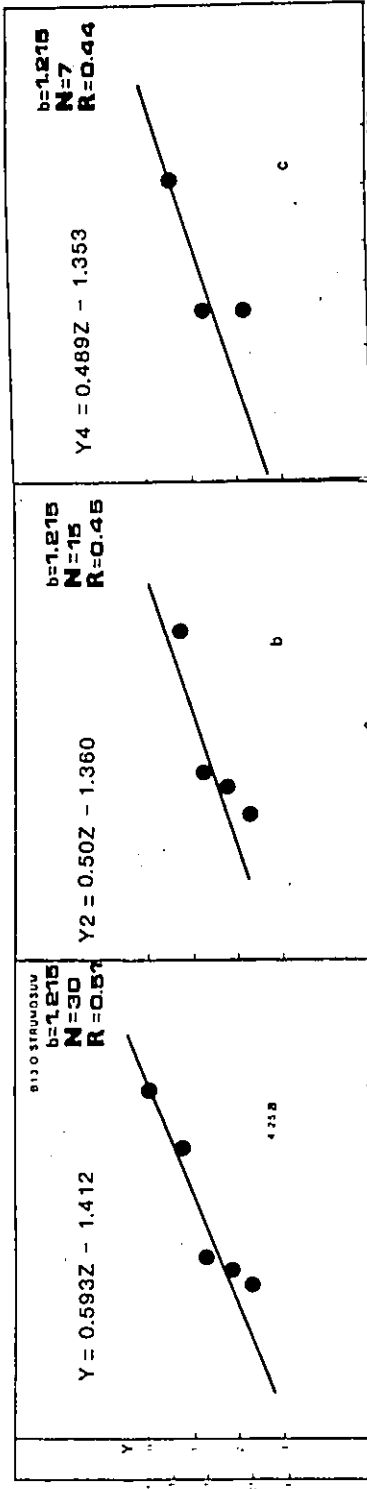
The first set of 3 plots is for B1 (Eoguttulina sp.). Eoguttulina sp. has mean a equal to 0.02847 and sample variances equal to 0.00020 0.00016 and 0.00011 for 7, 15 and 30 samples respectively. The second set of 3 plots is for B2 (Epistomina mosquensis). Epistomina mosquensis has a mean equal to 0.13840 and sample variances equal to 0.01236, 0.01038 and 0.00871. As shown in Fig 8.13, the third set of 3 plots is for B11 (Spirillina tenuissima), the fourth set of 3 plots is for B13 (Ophthalmidium stromosum) and the fifth set of 3 plots is for B14 (agglutinants). Their means, variances, b's and R's values were also estimated for 7, 15, and 30 samples, respectively.

The second set of data consisted of Gradstein's data. This data set included seventeen samples from Tojeira 1 and thirteen samples from Tojeira 2. In general, the ARIMA process requires a minimum of 30 samples for reliable results. Since each section had less than 30 samples the results of the ARIMA process on Gradstein's data must be considered with caution.

Third, an enlarged data set, which consisted of both Stam's and Gradstein's data sets, was applied to 14 species for both sections.

Figure 8.13 Experimental frequency distributions and their theoretical distributions for 30, 15 and 7 values of Eoquittulina sp., E. mosquensis, S. tenuissima, O. strumosum and Agglutinants for Tojeira 2 section with Stam's data





Nine examples as previously are explained for the Tojeira 1 section using Stam's data only are now used for a graphical test of the model ( Fig. 8.14 ) using Stam's and Gradstein's data together. As before, the first set of 3 plots is for A2( Eoguttulina sp.). They now have an average value equal to 0.02202 and variances equal to 0.00163, 0.00130 and 0.00104 for 10, 20 and 41 samples, respectively. The new results can be seen in Fig. 8.14. The second set of 3 plots is for A3 ( Epistomina mosquensis ) and the third set of 3 plots are for A10 ( Ophthalmidium strumosum )

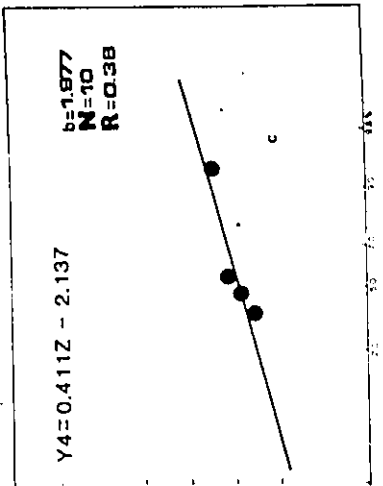
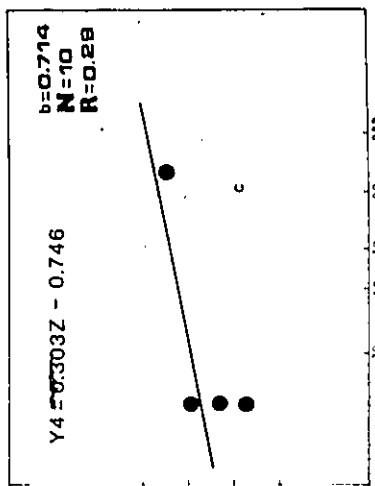
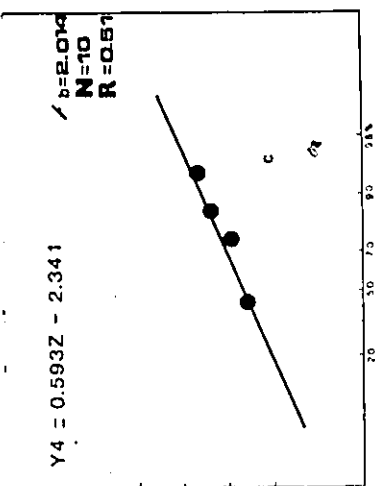
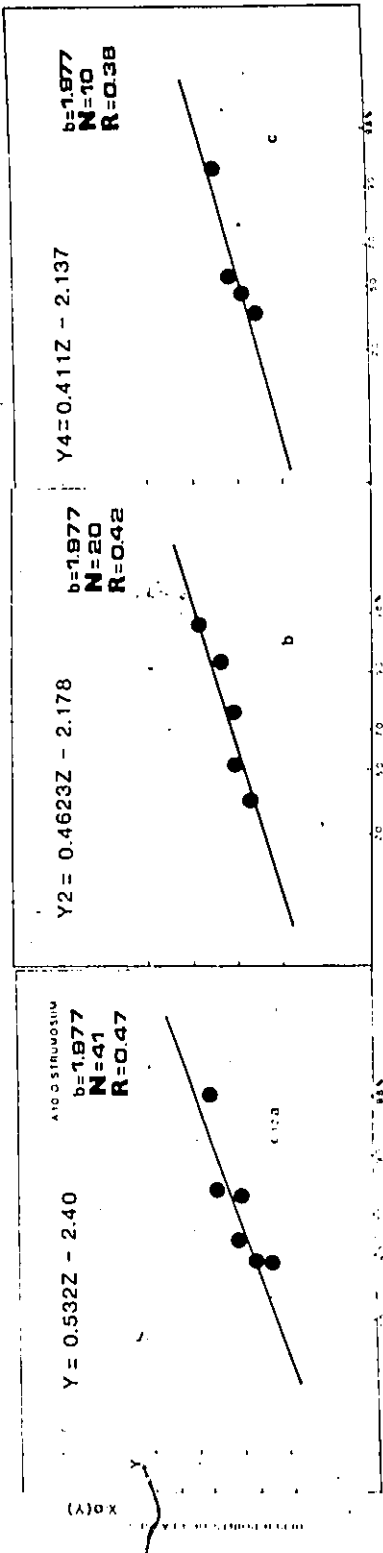
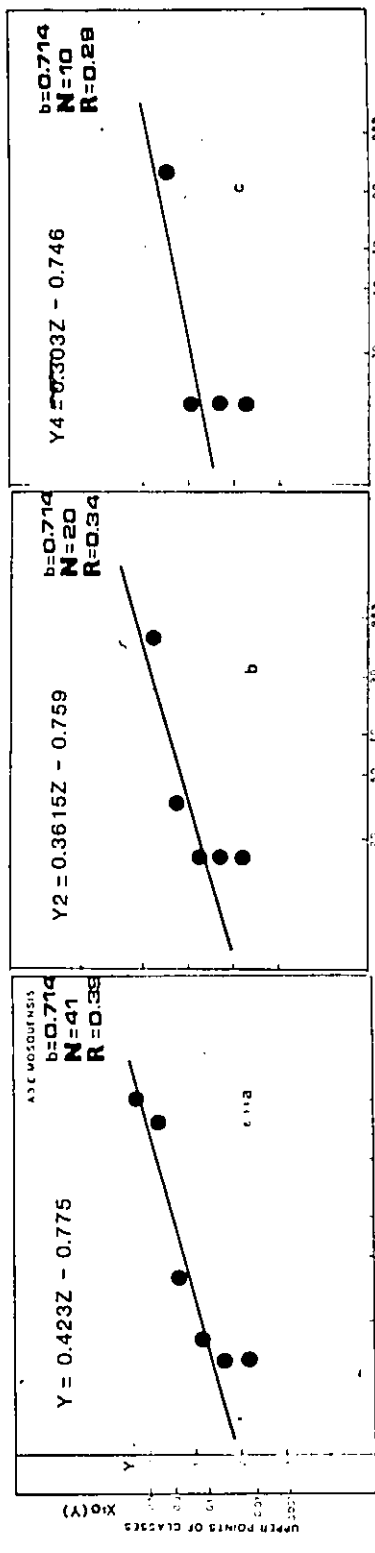
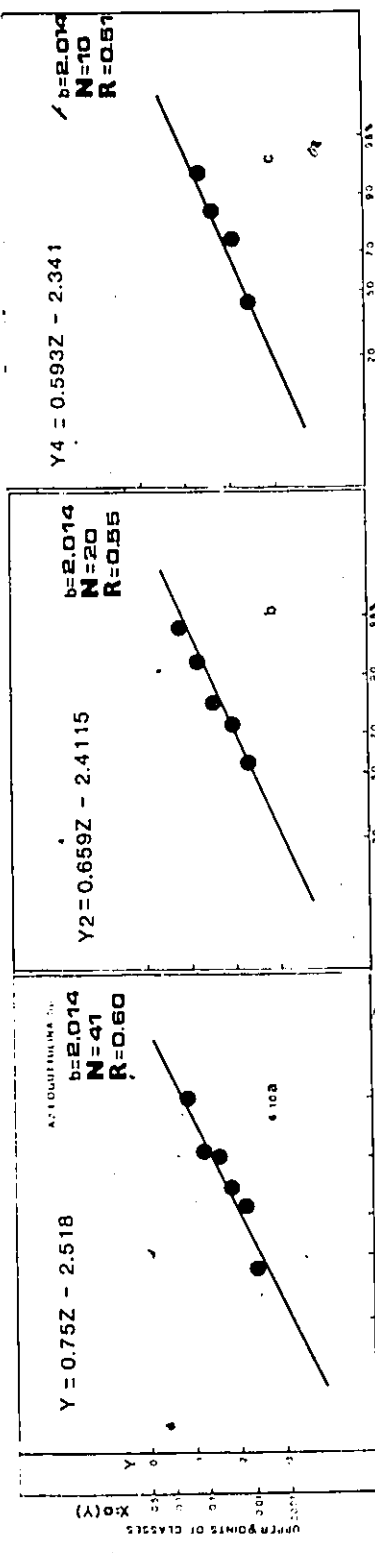
It may be concluded that the model given by equation ( 60 ) provides a good fit to the data used in Fig. 8.12, 8.13 and 8.14. The observed data tend to lie about the line approximated by the theoretical line described by Eq. 60 with  $b$  and  $R$  estimated from the means and variances of the proportion values.

The  $R$  values were tabulated with respect to their means and variances, their logarithmic means and variances or their means and quantities equal to ( variance / ( mean ( 1.0 - mean ) ) ) ( see appendix II ).  $R$  values can also be estimated from plots that were constructed according to mean and variance relationships ( Appendix III ).

The results shown in Figs. 8.12, 8.13 and 8.14 indicate that the new model provides a good fit for all sample sizes. Furthermore, the model given by Eq. 60 remains applicable with decreasing sample size.

The Tojeira 1 and Tojeira 2 sections were correlated with each other using Stam's data. Before correlation, the raw data were transformed with a logistical transformation. Positive correlation

Figure 8.14 Experimental frequency distributions and their theoretical distributions ( probnormal ) for 41, 20 and 10 values of Eoguttulina sp., E. mosquensis and O. strumosum with enlarged data



does not mean that the taxa and biological environments are dependent on each other. It means that they have similar relations with the same set of environmental conditions.

Epistomina mosquensis, Ophthalmidium strumosum and agglutinants show a positive correlation ( Table 7.1 and 7.2 ).

Epistomina mosquensis has been reported from deeper shelf deposits ( Pazdro, 1969 ). Epistomina mosquensis shows negative correlation with Eoguttulina sp. and with Spirillina tenuissima which prefer shallow water ( Stam, 1986).

Eoguttulina sp., Epistomina mosquensis, Ophthalmidium strumosum Spirillina tenuissima and agglutinants are unique species. The distribution of these taxa could be of bistratigraphic values. they could be used for correlation between sections using the abundance data for this species as well as for correlation with other species within the same samples.

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APPENDIX I : SAS and FORTRAN programs are used in this project. Listing of FORTRAN programs

1 - FF FORTRAN

2 - F FORTRAN

3 - ARIMA for Tojeira 1 and 2 sections with Stam's data

4 - ARIMA for Tojeira 1 and 2 sections with Gradstein's data

5 - ARIMA for Tojeira 1 section with enlarged data

1 - FF FORTRAN

```

PROGRAM HERMIT
DO 100 K=1,11
READ 1,A,G,X
FORNAT (F6.4,F10.8,F6.4)
IF (G.EQ.0.0) GO TO 999
DO 10 I= 1,70
RI=FLOAT (I)-1.0
RI=RI/100.0
R=0.3+RI
SMR=R**2+R**4**2/2.0
SMR=SMR+R**6*(A**2-1.0)**2/5.0
SMR=SMR+R**8*(A**3-3.0*A)**2/24.0
SMR=SMR+R**10*(A**4-6.0*A**2+3.0)**2/120.0
SMR=SMR+R**12*(A**5-10.0*A**3+15.0*A)**2/720.0
SMR=SMR+R**14*(A**6-15.0*A**4+45.0*A**2-15.0)**2/5040.0
SMR=SMR+R**16*(A**7-21.0*A**5+105.0*A**3-105.0*A)**2/40320.0
SMR9=R**18*(A**8-28.0*A**6+210.0*A**4-420.0*A**2+105.0)**2
SMR9=SMR9/362880.0+SMR
SMR10=A**9-36.0*A**7+378.0*A**5-1260.0*A**3+945.0*A
SMR10=SMR10**2
SMR10=SMR10/3628800.0+SMR9
SMR11=A**10-45.0*A**8+630.0*A**6-3150.0*A**4+4725.0*A**2-945.0
SMR11=SMR11**2
SMR11=R**22*SMR11
SMR11=SMR11/39916800.0
SMR11=SMR11+SMR10
SMR12=A**11-55.0*A**9+990.0*A**7-6930.0*A**5+17325.0*A**3-
+10395.0*A
SMR12=R**24*SMR12
SMR12=SMR12/479000000.0+SMR11
SMR13=A**12-66.0*A**10+1485.0*A**8-13860.0*A**6+51975.0*A**4-
+62370.0*A**2+10395.0
SMR13=SMR13**2
SMR13=R**26*SMR13
SMR13=SMR13/6227000000.0+SMR12
SMR11=SMR11+G**2
SMR12=SMR12+G**2
SMR13=SMR13+G**2
Y=SMR12/(X*(1.0-X))
100 PRINT 110,A,R,SMR11,SMR12,SMR13,Y
110 FORMAT (2(F6.4,5X),4(F10.0,5X))
100 CONTINUE
999 STOP
END

```

```

FF 00010
FF 00020
FF 00030
FF 00040
FF 00050
FF 00060
FF 00070
FF 00080
FF 00090
FF 00100
FF 00110
FF 00120
FF 00130
FF 00140
FF 00150
FF 00160
FF 00170
FF 00180
FF 00190
FF 00200
FF 00210
FF 00220
FF 00230
FF 00240
FF 00250
FF 00260
FF 00270
FF 00280
FF 00290
FF 00300
FF 00310
FF 00320
FF 00330
FF 00340
FF 00350
FF 00360
FF 00370
FF 00380
FF 00390
FF 00400
FF 00410
FF 00420
FF 00430
FF 00440
FF 00450

```

2 - F FORTRAN

```

DIMENSION X(50),A1(50),B2(50),C3(50),D4(50),E5(50),F6(50),H7(50)
DIMENSION P8(50),R9(50),A11(50),B22(50),C33(50),D44(50),E55(50)
DIMENSION F66(50),H77(50)
DO 110 MM=1,3
105 READ(5,105) N
FORMAT(I2)
READ(5,103) (X(I),I=1,N)
103 FORMAT(13F4.1)
SUMX=0.0
SUMXX=0.0
SUMAM=0.0
SUMAA=0.0
SUMBN=0.0
SUMBB=0.0
SUMCT=0.0
SUMCC=0.0
SUMDF=0.0
SUMDD=0.0
SUMEX=0.0
SUMEE=0.0
SUMFZ=0.0
SUMFF=0.0
SUMHW=0.0
SUMHH=0.0
DO 102 I=1,N
SUMX=SUMX+X(I)
SUMXX=SUMXX+X(I)**2
102 CONTINUE
DO 101 K=1,N-1
A1(K)=X(K)+X(K+1)
A11(K)=(A1(K))/2.0
SUMAM=SUMAM+A11(K)
SUMAA=SUMAA+A11(K)**2
101 CONTINUE
DO 100 L=1,N-2
B2(L)=A1(L)+X(L+2)
B22(L)=(B2(L))/3.0
SUMBN=SUMBN+B22(L)
SUMBB=SUMBB+B22(L)**2
100 CONTINUE
DO 99 M=1,N-3
C3(M)=B2(M)+X(M+3)
C33(M)=(B2(M)+X(M+3))/4.0
SUMCT=SUMCT+C33(M)
SUMCC=SUMCC+C33(M)**2
99 CONTINUE
DO 98 J=1,N-4
D4(J)=C3(J)+X(J+4)
D44(J)=D4(J)/5.0
SUMDF=SUMDF+D44(J)
SUMDD=SUMDD+D44(J)**2
98 CONTINUE
DO 97 NN=1,N-5
E5(NN)=D4(NN)+X(NN+5)
E55(NN)=E5(NN)/6.0
SUMEX=SUMEX+E55(NN)
SUMEE=SUMEE+E55(NN)**2
97 CONTINUE
DO 96 JJ=1,N-6
F6(JJ)=E5(JJ)+X(JJ+6)
F66(JJ)=F6(JJ)/7.0
SUMFZ=SUMFZ+F66(JJ)
SUMFF=SUMFF+F66(JJ)**2
96 CONTINUE
S0=(SUMXX-(SUMX**2/FLOAT(N)))/FLOAT(N-1)
S1=(SUMAA-(SUMAM**2/FLOAT(N-1)))/FLOAT(N-2)
S2=(SUMBB-(SUMBN**2/FLOAT(N-2)))/FLOAT(N-3)
S3=(SUMCC-(SUMCT**2/FLOAT(N-3)))/FLOAT(N-4)
S4=(SUMDD-(SUMDF**2/FLOAT(N-4)))/FLOAT(N-5)
S5=(SUMEE-(SUMEX**2/FLOAT(N-5)))/FLOAT(N-6)
S6=(SUMFF-(SUMFZ**2/FLOAT(N-6)))/FLOAT(N-7)
WRITE(6,*) S0,S1,S2,S3,S4,S5,S6
110 CONTINUE
STOP
END

```

```

F 00010
F 00020
F 00030
F 00040
F 00050
F 00060
F 00070
F 00080
F 00090
F 00100
F 00110
F 00120
F 00130
F 00140
F 00150
F 00160
F 00170
F 00180
F 00190
F 00200
F 00210
F 00220
F 00230
F 00240
F 00250
F 00260
F 00270
F 00280
F 00290
F 00300
F 00310
F 00320
F 00330
F 00340
F 00350
F 00360
F 00370
F 00380
F 00390
F 00400
F 00410
F 00420
F 00430
F 00440
F 00450
F 00460
F 00470
F 00480
F 00490
F 00500
F 00510
F 00520
F 00530
F 00540
F 00550
F 00560
F 00570
F 00580
F 00590
F 00600
F 00610
F 00620
F 00630
F 00640
F 00650
F 00660
F 00670
F 00680
F 00690
F 00700
F 00710
F 00720
F 00730
F 00740
F 00750

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3 - ARIMA for Tojeira 1 and 2 sections with Stam's data

SAS

12:42 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE = A1  
 MEAN OF WORKING SERIES = 6.47097  
 STANDARD DEVIATION = 4.66713  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	21.7821	1.00000												*****										0.179605
1	-1.72983	-0.07942											**											0.180734
2	3.91989	0.17996												****										0.186425
3	1.36552	0.06269												*										0.187104
4	1.39608	0.06409												*										0.187811
5	0.476232	0.02186												*										0.187893
6	1.31162	0.06022												*										0.188514
7	0.798186	0.03664												*										0.188744
8	1.1767	0.05402												*										0.189242
9	1.07484	0.04935												*										0.189657
10	-1.22516	-0.05625											*											

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.08837											**											
2	-0.16292										***	**											
3	-0.08739										**	*											
4	-0.02719										*	*											
5	0.01360										*	*											
6	-0.02652										*	*											
7	-0.04160										*	*											
8	-0.06838										*	*											
9	-0.02290										*	*											
10	0.07444										*	*											

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.07942											**											
2	0.17476												***										
3	0.09172												***										
4	0.04602												*										
5	0.00317												*										
6	0.03885												*										
7	0.03329												*										
8	0.03971												*										
9	0.03825												*										
10	-0.07885											**											

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	1.82	6	0.935	-0.079	0.180	0.063	0.064	0.022	0.060

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ARIMA PROCEDURE

NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES = 2.77097  
 STANDARD DEVIATION = 4.48713  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	20.1343	1.00000												*****										0
1	9.83615	0.48853												*****										0.179605
2	4.60299	0.22861												****										0.218301
3	6.48433	0.32205												*****										0.225892
4	5.80549	0.28834												*****										0.240247
5	1.74721	0.08678												**										0.251162
6	-0.0175513	-0.00087																						0.252128
7	-0.106569	-0.00529																						0.252128
8	-0.377554	-0.01875																						0.252131
9	-0.502992	-0.02498																						0.252176
10	-0.975797	-0.04846												*										0.252256

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.39887											*****											
2	-0.15046												***										
3	-0.17302												**										
4	-0.12338												*										
5	0.05685												*										
6	0.03231												*										
7	0.02655												*										
8	0.03444												*										
9	-0.04236												*										
10	0.01083												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.48853												*****										
2	-0.01319												*****										
3	0.28289												*****										
4	0.03506												*****										
5	-0.12060												**										
6	-0.07626												**										
7	-0.06344												*										
8	0.00149												*										
9	0.04306												*										
10	-0.01458												*										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS						
LAG	6	17.22	6	0.009	0.489	0.229	0.322	0.288	0.087	-0.001	12:42 MONDAY, OCTOBER 5, 1987

SAS

ARIMA PROCEDURE

NAME OF VARIABLE - A3  
 MEAN OF WORKING SERIES - 22.471  
 STANDARD DEVIATION - 12.6522  
 NUMBER OF OBSERVATIONS - 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	160.079	1.00000												*****											0	0.179605
1	79.9485	0.49343												*****											0	0.219887
2	85.2347	0.53245												*****											0	0.25815
3	58.3794	0.36469												*****											0	0.274266
4	32.2471	0.20145												*****											0	0.278998
5	27.9955	0.17489												*****											0	0.282512
6	14.9058	0.09312												*****											0	0.283501
7	25.9934	0.16238												*****											0	0.286485
8	23.4033	0.14620												*****											0	0.288882
9	19.8307	0.12388												*****											0	0.288882
10	12.4919	0.07804												*****											0	0.29059

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.05880												*										
2	-0.39872												*****										
3	-0.21208												****										
4	0.10481												****	**									
5	0.14240												****	***									
6	0.13767												****	***									
7	-0.12063												****	**									
8	-0.13749												****	**									
9	-0.00195												****	**									
10	0.08213												****	**									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.49943												*****										
2	0.37708												*****										
3	0.01664												*****										
4	-0.18030												****										
5	0.00757												****										
6	0.03097												****										
7	0.16038												****										
8	0.06029												****										
9	-0.06668												****										
10	-0.11801												****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	26.46	6	0.000	0.499	0.532	0.365	0.201	0.175	0.093

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ARIMA PROCEDURE

NAME OF VARIABLE - A4  
 MEAN OF WORKING SERIES- 3.36129  
 STANDARD DEVIATION - 3.70368  
 NUMBER OF OBSERVATIONS- 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	13.7172	1.00000												*****										0.179605
1	-0.0383418	-0.00280												*										0.179607
2	0.841339	0.06133												.										0.180281
3	-2.75041	-0.20051												****										0.187337
4	1.6674	0.12156												.										0.189864
5	3.96199	0.28883												****										0.203545
6	-2.44681	-0.17838												.										0.208527
7	-1.97001	-0.14362												****										0.211593
8	-3.54132	-0.25817												****										0.221617
9	1.22872	0.08958												.										0.222782
10	-1.91904	-0.13990												***										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.04815												*										
2	-0.23355												****										
3	0.30592												.										
4	0.01545												.										
5	-0.38451												*****										
6	0.14956												.										
7	0.19767												.										
8	0.00011												.										
9	0.02264												.										
10	0.18367												.										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.00280												*										
2	0.06133												.										
3	-0.20093												****										
4	0.12469												.										
5	0.32728												.										
6	-0.28702												*****										
7	-0.16683												.										
8	-0.08904												.										
9	-0.04150												.										
10	-0.26761												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB-	AUTOCORRELATIONS					
6	6.75	6	0.345	-0.003	0.061	-0.201	0.122	0.289	-0.178

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ARIMA PROCEDURE

NAME OF VARIABLE = AS  
 MEAN OF WORKING SERIES= 3.18065  
 STANDARD DEVIATION = 3.83426  
 NUMBER OF OBSERVATIONS= 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD	
0	14.7016	1.00000												*****											0
1	3.21369	0.21860												****											0.179605
2	-2.91795	-0.19848												*****											0.187992
3	-3.86038	-0.26460												*****											0.194634
4	-4.38163	-0.28443												*****											0.205746
5	-1.53857	-0.10465												**											0.218061
6	-0.892851	-0.06073												*											0.219676
7	-0.395641	-0.02691												*											0.220217
8	1.14899	0.07815												**											0.220323
9	0.118924	0.00809												*											0.221215
10	0.52068	0.03542												*											0.221225

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.46397												*****										
2	0.62334												*****										
3	0.51071												*****										
4	0.50650												*****										
5	0.35377												*****										
6	0.32722												*****										
7	0.25312												***										
8	0.13863												**										
9	0.12349												*										
10	0.09072												**										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.21860												****										
2	-0.25862												****										
3	-0.17046												***										
4	-0.26489												****										
5	-0.10707												**										
6	-0.23353												****										
7	-0.19589												****										
8	-0.13432												****										
9	-0.26045												****										
10	-0.19116												****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI																							
LAG	SQUARE	DF	PROB	0.219	-0.198	-0.263	-0.284	-0.105	-0.061															
6	9.19	6	0.163																					

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ARIMA PROCEDURE

NAME OF VARIABLE - AG  
 MEAN OF WORKING SERIES - 10.2419  
 STANDARD DEVIATION - 5.9591  
 NUMBER OF OBSERVATIONS - 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	35.5108	1.00000												*****											0	0.179605
1	7.74813	0.21819												****											0	0.187961
2	-2.21072	-0.06225												*****											0	0.188625
3	-12.2193	-0.34410												*****											0	0.207891
4	1.7834	0.05022												*****											0	0.208282
5	7.55795	0.21284												*****											0	0.215183
6	-2.76041	-0.07773												*****											0	0.216087
7	-9.98317	-0.28113												*****											0	0.22758
8	-3.3978	-0.09568												*****											0	0.228874
9	3.65937	0.10305												*****											0	0.230366
10	9.38114	0.26418												*****												

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.29908												*****										
2	-0.38702												*****										
3	0.54935												*****										
4	-0.02219												*****										
5	-0.36353												*****										
6	0.26237												*****										
7	0.10517												*****										
8	-0.16590												*****										
9	0.07907												*****										
10	-0.01973												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.21819												*****										
2	-0.11535												*****										
3	-0.32327												*****										
4	0.22505												*****										
5	0.14637												*****										
6	-0.35507												*****										
7	-0.13750												*****										
8	0.25738												*****										
9	-0.12726												*****										
10	0.03779												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	8.21	6	0.223	0.218	-0.062	-0.344	0.050	0.213	-0.078

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ARIMA PROCEDURE

NAME OF VARIABLE = A7  
 MEAN OF WORKING-SERIES = 4.73548

STANDARD DEVIATION = 2.81477  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	7.92293	1.00000												***										0.179605
1	1.10392	-0.13933												***										0.183059
2	-0.0276046	-0.00348												***										0.183061
3	1.22344	0.15442												***										0.187216
4	-1.09178	-0.13780												***										0.190459
5	0.754833	0.09527												***										0.191991
6	-0.727153	-0.09178												***										0.193401
7	-0.841242	-0.10618												***										0.195272
8	-0.607848	-0.07672												***										0.196242
9	-1.96482	-0.24799												***										0.206103
10	-1.67154	-0.21097												***										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.38413												***										
2	0.40548												***										
3	-0.37965												***										
4	0.37481												***										
5	-0.30921												***										
6	0.25903												***										
7	-0.14434												***										
8	0.15599												***										
9	0.00262												***										
10	0.16411												***										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.13933												***										
2	-0.02335												***										
3	0.16139												***										
4	-0.19177												***										
5	0.16925												***										
6	-0.20026												***										
7	0.03173												***										
8	-0.18538												***										
9	-0.11470												***										
10	-0.25130												***										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	2.95	6	0.814	0.139	-0.003	0.154	-0.138	0.095	-0.092

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ARIMA PROCEDURE

NAME OF VARIABLE = AB  
 MEAN OF WORKING SERIES = 2.13871  
 STANDARD DEVIATION = 1.28557  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD	
0	1.6527	1.00000												*****											0
1	0.679098	0.41090												*****											0.179605
2	0.107905	0.06529												*****											0.207728
3	-0.36514	-0.22094												*****											0.208389
4	-0.119194	-0.07212												*****											0.215813
5	-0.0759129	-0.04593												*****											0.216589
6	0.0579618	0.03507												*****											0.216903
7	0.322502	0.19514												*****											0.217086
8	0.265222	0.16048												*****											0.222673
9	0.0828116	0.05011												*****											0.226373
10	-0.298475	-0.18060												*****											0.22673

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.42496												*****										
2	-0.10749												*****										
3	0.37282												*****										
4	-0.27712												*****										
5	0.04671												*****										
6	0.13220												*****										
7	-0.16498												*****										
8	0.04037												*****										
9	-0.04214												*****										
10	0.06562												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.41090												*****										
2	-0.12459												*****										
3	-0.24432												*****										
4	0.16168												*****										
5	-0.08013												*****										
6	0.06688												*****										
7	0.26953												*****										
8	-0.07033												*****										
9	-0.03590												*****										
10	-0.10181												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS				
6	8.02	6	0.237	0.411	0.065	-0.221	-0.072	-0.046	0.035

SAS 12:42 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE = A9  
 MEAN OF WORKING SERIES = 3.16774  
 STANDARD DEVIATION = 6.59889  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	43.5454	1.00000												*****											0	0.179605
1	2.56896	0.05899												*											0	0.180229
2	1.32893	0.03052												*											0	0.180396
3	11.1393	0.25581												****											0	0.191741
4	-6.02295	-0.13831												*											0	0.194933
5	2.81595	0.06467												*											0	0.195623
6	-1.75948	-0.04041												*											0	0.195893
7	-4.63903	-0.10653												****											0	0.197753
8	7.75166	0.17801												*											0	0.202856
9	-0.0426226	-0.00098												*											0	0.202856
10	-1.93053	-0.04433												*												

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.15858												***										
2	0.04762												*										
3	-0.28828												*****										
4	0.13699												*										
5	-0.04629												*										
6	0.13468												***										
7	0.01087												*										
8	-0.11214												**										
9	-0.03875												*										
10	0.04062												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.05899												*										
2	0.02713												*										
3	0.25352												*****										
4	-0.17077												****										
5	0.08499												*										
6	-0.12817												***										
7	-0.00705												*										
8	0.14075												***										
9	0.03408												*										
10	-0.04747												*										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS				
6	3.50	6	0.744	0.059	0.031	0.256	-0.138	0.065	-0.040

SAS 12:42 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE = A10  
 MEAN OF WORKING SERIES= 1.92903  
 STANDARD DEVIATION = 2.44503  
 NUMBER OF OBSERVATIONS= 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	5.97819	1.00000																						0.179605
1	2.42279	0.40527												*****										0.207013
2	1.38282	0.19786												****										0.213026
3	2.97305	0.49732												*****										0.247662
4	1.9424	0.32491												*****										0.261051
5	0.788303	0.13186												***										0.26319
6	1.52585	0.25524												*****										0.271057
7	1.10028	0.18405												***										0.275059
8	-0.0387607	-0.00648												**										0.275064
9	0.736675	0.12323												*										0.276839
10	0.282735	0.04729												*										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.41223												*****										
2	0.28358												*****										
3	-0.34462												*****										
4	0.04717												*****										
5	-0.10078												*****										
6	0.09322												*****										
7	-0.15191												*****										
8	0.17082												*****										
9	-0.13653												*****										
10	0.09781												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.40527												*****										
2	0.04022												*	*****									
3	0.48425												*	*	*****								
4	-0.04037												*	*	*	*****							
5	-0.00047												*	*	*	*	*****						
6	0.02176												*	*	*	*	*	*****					
7	-0.06657												*	*	*	*	*	*	*****				
8	-0.11777												*	*	*	*	*	*	*	*****			
9	0.08772												*	*	*	*	*	*	*	*	*****		
10	-0.14693												*	*	*	*	*	*	*	*	*	*****	

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	23.37	6	0.001	0.405 0.198 0.497 0.325 0.132 0.255

ARIMA PROCEDURE

NAME OF VARIABLE = All  
 MEAN OF WORKING SERIES = 7.80968  
 STANDARD DEVIATION = 7.16498  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	51.337	1.00000												*****										0
1	26.3863	0.51398												*****										0.179605
2	13.0168	0.25356												*****										0.22204
3	9.56743	0.18637												*****										0.231192
4	-3.5746	-0.06943												*****										0.235988
5	-7.5117	-0.14632												*****										0.23665
6	-2.3727	-0.04622												*****										0.23955
7	-1.99501	-0.03886												*****										0.239838
8	-7.45035	-0.14513												*****										0.240041
9	-7.73027	-0.15058												*****										0.242855
10	-3.14883	-0.06134												*****										0.245848

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.38910												*****										
2	0.00328												*****										
3	-0.24165												*****										
4	0.24735												*****										
5	0.08416												*****										
6	-0.08535												*****										
7	-0.10398												*****										
8	0.04695												*****										
9	0.13012												*****										
10	-0.06895												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.51398												*****										
2	-0.01444												*****										
3	0.08371												*****										
4	-0.26738												*****										
5	-0.02047												*****										
6	0.09151												*****										
7	-0.01603												*****										
8	-0.19356												*****										
9	-0.08524												*****										
10	0.10330												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS																			
6	13.66	6	0.034	0.514	0.254	0.186	-0.070	-0.146	-0.046														

ARIMA PROCEDURE

NAME OF VARIABLE = A12  
 MEAN OF WORKING SERIES = 3.91613  
 STANDARD DEVIATION = 2.92928  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	8.58071	1.00000												*****										0.179605
1	4.29819	0.50091												***										0.220105
2	1.17676	0.13714												**										0.222844
3	-0.962835	-0.11211												*****										0.224659
4	-2.43778	-0.28410												*****										0.235964
5	-3.16228	-0.36853												*****										0.253853
6	-2.8985	-0.33779												*****										0.267961
7	-2.55518	-0.29778												*										0.278431
8	-0.618964	-0.07211												*										0.278431
9	0.143681	0.01674												*										0.279033
10	-0.374912	-0.04369												*										0.279066

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.26717												*****										
2	0.09241												**										
3	0.15864												***										
4	0.08799												**										
5	-0.16314												***										
6	-0.00021												***										
7	0.24329												*****										
8	-0.04734												*										
9	-0.06288												*										
10	0.19272												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.50091												*****										
2	-0.15188												***										
3	-0.15750												***										
4	-0.18886												***										
5	-0.18326												***										
6	-0.12623												***										
7	-0.19568												***										
8	0.06038												*										
9	-0.15329												***										
10	-0.27604												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	0.501	AUTOCORRELATIONS	0.137	-0.112	-0.284	-0.369	-0.338	12:42 MONDAY, OCTOBER 5, 1987
6	22.75	6	0.001	0.501	SAS						

ARIMA PROCEDURE

NAME OF VARIABLE = A13  
 MEAN OF WORKING SERIES = 2.16452  
 STANDARD DEVIATION = 5.7693  
 NUMBER OF OBSERVATIONS = 31  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	33.2849	1.00000												*****										0
1	14.6964	0.44154												*****										0.179605
2	4.73734	0.14233												***										0.211744
3	7.90004	0.23735												*****										0.214808
4	5.20103	0.15626												***										0.223108
5	1.44456	0.04340												*										0.22661
6	-0.762471	-0.02291												.										0.226878
7	-1.17215	-0.03522												.										0.226953
8	-1.01493	-0.03049												.										0.227129
9	-1.01704	-0.03056												.										0.227261
10	-1.34978	-0.04055												.										0.227394

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.49257												*****										
2	0.27666												*****										
3	-0.27158												*****										
4	0.08508												**										
5	-0.07421												*										
6	0.07744												**										
7	-0.01609												.										
8	0.02684												.										
9	-0.02013												.										
10	0.01104												.										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.44154												*****										
2	-0.04537												*										
3	0.24858												*****										
4	-0.05725												*										
5	-0.00425												.										
6	-0.09391												**										
7	-0.01292												.										
8	-0.01486												.										
9	0.01273												.										
10	-0.01542												.										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS									
6	10.44	6	0.107	0.442	0.142	0.237	0.156	0.043	-0.023					

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ARIMA PROCEDURE

NAME OF VARIABLE = A14  
 MEAN OF WORKING SERIES= 16.9581  
 STANDARD DEVIATION = 8.85819  
 NUMBER OF OBSERVATIONS= 31

AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	78.4676	1.00000												*****										0
1	11.3129	0.14417												***										0.179605
2	-11.1688	-0.14234																						0.183301
3	18.3317	0.23362																						0.186832
4	33.6093	0.42832																						0.196029
5	2.83115	0.03608												*										0.224195
6	-1.99423	-0.02541												*										0.224382
7	13.2224	0.16851												***										0.224175
8	1.55163	0.01977												**										0.228519
9	8.10947	0.10335												**										0.228574
10	9.15769	0.11671												**										0.230077

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.11838												**										
2	0.06853												*										
3	-0.10076												**										
4	-0.38181												*****										
5	0.03583												*										
6	0.09942												**										
7	-0.06439												*										
8	0.17800												*****										
9	-0.12390												**										
10	-0.02677												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.14417												***										
2	-0.16658												***										
3	0.29576												*****										
4	0.35002												*****										
5	0.00122												*										
6	0.03817												*										
7	-0.00334												*										
8	-0.20574												****										
9	0.17752												****										
10	0.03608												*										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	DF	PROB	AUTOCORRELATIONS					
LAG	SQUARE								
6	10.45	6	0.107	0.144	-0.142	0.234	0.428	0.036	-0.025

SAS

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ARIMA PROCEDURE

NAME OF VARIABLE - B1  
 MEAN OF WORKING SERIES - 2.84657  
 STANDARD DEVIATION - 1.43079  
 NUMBER OF OBSERVATIONS - 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	2.04716	1.00000												*****										0
1	0.952839	0.46545												*****										0.182574
2	0.638966	0.31212												*****										0.218577
3	0.440271	0.21506												*****										0.232961
4	0.728687	0.35595												*****										0.239487
5	0.721859	0.35262												*****										0.256517
6	0.564276	0.27564												*****										0.272195
7	-0.148975	-0.07277												*****										0.281345
8	0.178041	0.08697												**										0.281972
9	0.01608	0.00785												*										0.282865
10	0.0867185	0.04236												*										0.282872

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.35597												*****										
2	-0.07553												*****										
3	0.45279												*****										
4	-0.05504												*****										
5	-0.12524												*****										
6	-0.20078												*****										
7	0.33761												*****										
8	-0.19931												*****										
9	0.07789												*****										
10	0.03349												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.46545												*****										
2	0.12189												*****										
3	0.03986												*****										
4	0.28239												*****										
5	0.12452												*****										
6	0.00929												*****										
7	-0.37665												*****										
8	0.17039												*****										
9	-0.16726												*****										
10	-0.05192												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	'AUTOCORRELATIONS					
6	24.85	6	0.000	0.465	0.312	0.215	0.356	0.353	0.276

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ARIMA PROCEDURE

NAME OF VARIABLE = B2  
 MEAN OF WORKING SERIES = 13.84  
 STANDARD DEVIATION = 11.1162  
 NUMBER OF OBSERVATIONS = 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	123.57	1.00000																						0
1	82.6732	0.66904												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.182574
2	73.2918	0.59312												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.251345
3	65.7658	0.53211												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.294324
4	44.4355	0.35960												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.324823
5	36.0074	0.29139												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.337832
6	31.8627	0.25542												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.346109
7	19.0184	0.15391												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.352336
8	16.5723	0.13411												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.35457
9	9.96005	0.08060												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.356257
10	-3.74567	-0.03031												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	0.356864

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.24444												*****										
2	-0.16458											*****	*****										
3	-0.18472											*****	*****	*****									
4	0.08161											*****	*****	*****	*****								
5	0.07836											*****	*****	*****	*****	*****							
6	-0.02635											*****	*****	*****	*****	*****	*****						
7	-0.02726											*****	*****	*****	*****	*****	*****	*****					
8	-0.07548											*****	*****	*****	*****	*****	*****	*****	*****				
9	-0.06415											*****	*****	*****	*****	*****	*****	*****	*****	*****			
10	0.10547											*****	*****	*****	*****	*****	*****	*****	*****	*****	*****		

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.66904												*****										
2	0.26341												*****	*****									
3	0.12389												*****	*****	*****								
4	-0.17430												*****	*****	*****	*****							
5	-0.02600												*****	*****	*****	*****	*****						
6	0.06205												*****	*****	*****	*****	*****	*****					
7	-0.05618												*****	*****	*****	*****	*****	*****	*****				
8	0.01633												*****	*****	*****	*****	*****	*****	*****	*****			
9	-0.05145												*****	*****	*****	*****	*****	*****	*****	*****	*****		
10	-0.14598												*****	*****	*****	*****	*****	*****	*****	*****	*****		

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	47.59	6	0.000	0.669 0.593 0.532 0.360 0.291 0.255

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ARIMA PROCEDURE

NAME OF VARIABLE - B3  
 MEAN OF WORKING SERIES - 3.92667  
 STANDARD DEVIATION - 6.01049  
 NUMBER OF OBSERVATIONS - 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	36.126	1.00000												*	*	*	*	*	*	*	*	*	*	0
1	1.51753	0.04201										**	*											0.182574
2	-3.53858	-0.09795										**	*											0.182896
3	0.127884	0.00354										*	*											0.184636
4	-2.46216	-0.06815										*	*											0.184639
5	1.63624	0.04529										*	*											0.185475
6	1.5387	0.04259										**	*											0.185844
7	-4.5629	-0.12631										**	*											0.186169
8	-4.00759	-0.11093										**	*											0.189004
9	1.83885	0.05090										*	*											0.191162
10	4.40425	0.12191										*	*											0.191613

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.04220										*	*											
2	0.06401									*	*	*											
3	-0.03641									*	*	*	*										
4	0.09306									*	*	*	*	*									
5	-0.03051									*	*	*	*	*	*								
6	-0.03120									*	*	*	*	*	*	*							
7	0.10229									*	*	*	*	*	*	*	*						
8	0.08175									*	*	*	*	*	*	*	*	*					
9	-0.03291									*	*	*	*	*	*	*	*	*	*				
10	-0.10205									*	*	*	*	*	*	*	*	*	*	*			

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.04201										*	*											
2	-0.09989									*	*	*											
3	0.01241									*	*	*	*										
4	-0.07963									*	*	*	*	*									
5	0.05490									*	*	*	*	*	*								
6	0.02278									*	*	*	*	*	*	*							
7	-0.12017									*	*	*	*	*	*	*	*						
8	-0.10147									*	*	*	*	*	*	*	*	*					
9	0.04419									*	*	*	*	*	*	*	*	*	*				
10	0.10754									*	*	*	*	*	*	*	*	*	*	*			

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	0.71	6	0.994	0.042	-0.098	0.004	-0.068	0.045	0.043

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ARIMA PROCEDURE

NAME OF VARIABLE - B4

MEAN OF WORKING SERIES- 0.78  
 STANDARD DEVIATION - 1.1791  
 NUMBER OF OBSERVATIONS- 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	1.39027	1.00000																						0
1	0.219053	0.15756												***										0.182574
2	0.0451067	0.03244												*										0.187052
3	0.47116	0.33890												*****										0.187239
4	-0.00145333	-0.00105																						0.206677
5	-0.0312	-0.02244																						0.206677
6	0.175253	0.12606												***										0.206758
7	-0.00469333	-0.00338																						0.209305
8	-0.0341733	-0.02458																						0.209306
9	0.192547	0.13850												***										0.209403
10	0.0543333	0.03908												*										0.212434

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.23235											*****											
2	0.11859												**										
3	-0.33148											*****	**										
4	0.12301												*										
5	-0.04440												*										
6	0.05921												*										
7	-0.02410												*										
8	0.04745												*										
9	-0.08985											**											
10	-0.00517																						

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.15756											***											
2	0.00781												***										
3	0.34108												*	*****									
4	-0.12231											**		*									
5	0.00314													*									
6	0.01399													*									
7	0.01224													*									
8	-0.01454													*									
9	0.11015													*									
10	0.00620													*									

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	0.158	0.032	0.339	-0.001	-0.022	0.126
LAG	6	5.60	6	0.470	0.158	0.032	0.339	-0.001	-0.022	0.126

AUTOCORRELATIONS

SAS

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ARIMA PROCEDURE

NAME OF VARIABLE = B5  
 MEAN OF WORKING SERIES = 4.49667  
 STANDARD DEVIATION = 2.12783  
 NUMBER OF OBSERVATIONS = 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	4.52766	1.00000												*****											0	0.162574
1	0.709677	0.15674								*****															0	0.187006
2	-1.67178	-0.36924								*****															0	0.209905
3	-0.614223	-0.13566								*****															0	0.212808
4	0.292387	0.06458								*****															0	0.21346
5	0.139454	0.03080								*****															0	0.213608
6	0.213731	0.04721								*****															0	0.213956
7	-0.0601248	-0.01328								*****															0	0.213983
8	-0.643481	-0.14212								*****															0	0.217107
9	-0.522414	-0.11538								*****															0	0.219141
10	0.0617407	0.01364								*****																

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.21865									*****													
2	0.36645									*****													
3	0.02426									*****													
4	0.02148									*****													
5	0.08662									*****													
6	0.01198									*****													
7	0.05845									*****													
8	0.16023									*****													
9	0.05235									*****													
10	0.04799									*****													

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.15674									*****													
2	-0.40372									*****													
3	0.01085									*****													
4	-0.06937									*****													
5	-0.02711									*****													
6	0.06307									*****													
7	-0.05135									*****													
8	-0.11133									*****													
9	-0.09157									*****													
10	-0.06065									*****													

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	6.42	6	0.378	0.157	-0.369	-0.136	0.065	0.031	0.047

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ARIMA PROCEDURE

NAME OF VARIABLE = B6  
 MEAN OF WORKING SERIES = 3.90333  
 STANDARD DEVIATION = 2.06292  
 NUMBER OF OBSERVATIONS = 10  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	4.25566	1.00000																						0.182574
1	-0.552789	-0.12990										***												0.185629
2	-0.406812	-0.09559									**													0.187263
3	0.437443	0.10279									*													0.189134
4	0.085343	0.02005																						0.189205
5	0.817843	0.19218																						0.195604
6	-1.27698	-0.30007									*****													0.210389
7	0.236197	0.05550									*													0.210876
8	0.0830526	0.01952										**												0.210936
9	-0.385826	-0.09066										*												0.212231
10	-0.292304	-0.06869																						

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.01110												****										
2	0.17511										****												
3	-0.18919										*												
4	0.03787										****												
5	-0.21713										*		****										
6	0.22299										*		*										
7	-0.03498											*	*										
8	0.10951											*	*										
9	0.04786											*	*										
10	0.11592											*	*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.12990												***										
2	-0.11440											**											
3	0.07636											*											
4	0.03533											*											
5	0.22582											*											
6	-0.26738											*											
7	0.03030											*											
8	-0.08778											*											
9	-0.03499											*											
10	-0.13952											*											

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS					
LAG	6	6.28	6	0.392	-0.130	-0.096	0.103	0.020	0.192	-0.300

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SAS

ARIMA PROCEDURE

NAME OF VARIABLE - B7  
 MEAN OF WORKING SERIES- 2.75667  
 STANDARD DEVIATION - 1.57452  
 NUMBER OF OBSERVATIONS- 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	2.47912	1.00000	*****																							0
1	-0.511729	-0.20642	*****																							0.182574
2	1.02654	0.41407	*****																							0.190194
3	-0.437066	-0.17630	*****																							0.218184
4	0.537716	0.21690	*****																							0.222882
5	-0.248813	-0.10036	*****																							0.22981
6	0.31268	0.12613	*****																							0.231267
7	-0.870349	-0.35107	*****																							0.233548
8	-0.33489	-0.13508	*****																							0.250522
9	-0.209986	-0.08470	*****																							0.252939
10	-0.288159	-0.11623	*****																							0.253882

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.10318	*****																						
2	-0.37825	*****																						
3	0.02025	*****																						
4	0.03499	*****																						
5	-0.21618	*****																						
6	-0.11661	*****																						
7	0.36096	*****																						
8	0.19961	*****																						
9	-0.09582	*****																						
10	-0.00428	*****																						

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	-0.20642	*****																						
2	0.38800	*****																						
3	-0.05398	*****																						
4	0.03618	*****																						
5	0.01862	*****																						
6	0.01705	*****																						
7	-0.35606	*****																						
8	-0.37176	*****																						
9	0.14940	*****																						
10	0.00678	*****																						

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS					
LAG	6	11.15	6	0.084	-0.206	0.414	-0.176	0.217	-0.100	0.126

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ARIMA PROCEDURE

NAME OF VARIABLE - B8  
 MEAN OF WORKING SERIES - 2.73333  
 STANDARD DEVIATION - 3.06446  
 NUMBER OF OBSERVATIONS - 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	9.39089	1.00000												*****										0.182574
1	3.2033	0.34111												*****										0.202707
2	1.99281	0.21221												****										0.209982
3	2.60356	0.27724												****										0.221848
4	0.445519	0.04744												*										0.222186
5	0.984148	0.10486												**										0.223828
6	-0.473333	-0.05040												*										0.224206
7	0.308296	0.03283												*										0.224366
8	1.90059	0.20239												****										0.230371
9	-0.924556	-0.02845												**										0.231769
10	-0.733148	-0.07837												**										

MARK TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.29149												*****										
2	-0.01466												*****										
3	-0.22760												*****	**									
4	0.11758												*****	**									
5	-0.03351												*****	*									
6	0.15140												*****	*									
7	-0.03746												*****	*									
8	-0.22547												*****	*									
9	0.10198												*****	*									
10	0.06744												*****	*									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.34111												*****										
2	0.10847												*****										
3	0.20121												*****										
4	-0.13067												*****										
5	0.08424												*****										
6	-0.17821												*****										
7	0.13725												*****										
8	0.16688												*****										
9	-0.20288												*****										
10	-0.09083												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	0.341	0.212	0.277	0.047	0.105	-0.050
LAG	6	8.74	6	0.189						

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ARIMA PROCEDURE

NAME OF VARIABLE - B9  
 MEAN OF WORKING SERIES - 3.17  
 STANDARD DEVIATION - 1.47313  
 NUMBER OF OBSERVATIONS - 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	2.1701	1.00000												*****										0.182574
1	-0.29153	-0.13434										****												0.18584
2	-0.525627	-0.24221									*****													0.196081
3	0.0997767	0.04598									****		*											0.19644
4	-0.481953	-0.22204									****		****											0.204635
5	0.46615	0.21481									****		****											0.212018
6	0.34162	0.15742									****		****											0.215879
7	-0.267277	-0.12316									****		****											0.218208
8	-0.32724	-0.15079									****		****											0.221655
9	-0.0766367	-0.03531									****		****											0.221842
10	0.246933	0.11379									****		****											

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.12628												****										
2	0.28686												*****										
3	-0.02893												*****										
4	0.25037												*****										
5	-0.11637												*****										
6	0.00631												*****										
7	0.03441												*****										
8	0.13900												*****										
9	0.01242												*****										
10	-0.03098												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.13434												****										
2	-0.26504												*****										
3	-0.03373												*****										
4	-0.31135												*****										
5	0.15251												*****										
6	0.08178												*****										
7	0.03100												*****										
8	-0.19181												*****										
9	-0.02789												*****										
10	0.03916												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	7.27	6	0.297	-0.134	-0.242	0.046	-0.222	0.215	0.157

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ARIMA PROCEDURE

NAME OF VARIABLE = B10  
 MEAN OF WORKING SERIES = 1.75  
 STANDARD DEVIATION = 2.43143  
 NUMBER OF OBSERVATIONS = 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	STD
0	5.91183	1.00000	0
1	1.15208	0.19488	0.182574
2	1.47583	0.24964	0.189381
3	1.41358	0.23911	0.20005
4	0.314667	0.05323	0.20936
5	1.52342	0.25769	0.20981
6	0.593667	0.10042	0.220107
7	1.06442	0.18005	0.221629
8	0.916667	0.15506	0.226452
9	0.00891667	0.00151	0.229964
10	0.0388333	0.00657	0.229964

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION
1	-0.08506
2	-0.10941
3	-0.13372
4	0.09586
5	-0.14349
6	0.02797
7	-0.13122
8	-0.07424
9	0.04813
10	0.10112

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION
1	0.19488
2	0.22002
3	0.17291
4	-0.06315
5	0.19122
6	0.00800
7	0.10181
8	0.02725
9	-0.08498
10	-0.11571

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	8.48	6	0.205	0.195 0.250 0.239 0.053 0.258 0.100

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ARIMA PROCEDURE

NAME OF VARIABLE - B11  
 MEAN OF WORKING SERIES - 25.7533  
 STANDARD DEVIATION - 17.3525  
 NUMBER OF OBSERVATIONS - 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	301.11	1.00000																						0
1	203.829	0.67692												*****										0.182574
2	185.139	0.61485												*****										0.252748
3	175.348	0.58234												*****										0.29847
4	177.343	0.42291												*****										0.334204
5	142.935	0.47469												*****										0.351591
6	92.4682	0.30709												*****										0.372342
7	58.3148	0.30767												****										0.380691
8	28.4353	0.19367												**										0.383961
9	1.65364	0.00549												*										0.384734
10	-5.76119	-0.01913												*										0.384737

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.39549												*****										
2	0.13276												*****										
3	-0.31940												*****										
4	0.31233												*****										
5	-0.36177												*****										
6	0.09598												*****										
7	-0.07369												*****										
8	0.16188												*****										
9	-0.05475												*****										
10	0.02759												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.67692												*****										
2	0.28910												*****										
3	0.18280												*****										
4	-0.15411												*****										
5	0.29868												*****										
6	-0.23593												*****										
7	-0.12503												*****										
8	-0.22517												*****										
9	0.04611												*****										
10	-0.04606												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	AUTOCORRELATIONS							
LAG	SQUARE	DF	PROB	0.615	0.582	0.423	0.475	0.307	
6	59.22	6	0.000	0.677	0.615	0.582	0.423	0.475	0.307

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ARIMA PROCEDURE

NAME OF VARIABLE - B12  
 MEAN OF WORKING SERIES- 3.06667  
 STANDARD DEVIATION - 2.19884  
 NUMBER OF OBSERVATIONS- 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	4.83489	1.00000																						0
1	0.954185	0.19735												****										0.182574
2	2.13293	0.44115												*****										0.189552
3	0.564889	0.11684												**										0.221143
4	0.0462963	0.00958																						0.223191
5	-0.402852	-0.08332																						0.223205
6	-1.87056	-0.38689																						0.224239
7	-0.359481	-0.07435												*										0.245483
8	-1.37507	-0.28441												*****										0.246233
9	0.456444	0.09441												**										0.25695
10	-0.600593	-0.12422												**										0.258103

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.08776												**										
2	-0.34865												*****										
3	-0.37417												*****										
4	-0.10116												**										
5	0.13671													***									
6	0.36373													*****									
7	0.04028													*									
8	-0.11128													**									
9	-0.17617													*****									
10	0.08286													**									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.19735												****										
2	0.41850												*****										
3	-0.02067													*****									
4	-0.23528													***									
5	-0.13688													***									
6	-0.36804													*****									
7	0.15339													***									
8	0.10261													**									
9	0.27290													*****									
10	-0.13705													***									

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	0.197	0.441	0.117	0.010	-0.083	-0.387
6	14.70	6	0.023	0.197	0.441	0.117	0.010	-0.083	-0.387

ARIMA PROCEDURE

NAME OF VARIABLE = B13  
 MEAN OF WORKING SERIES = 11.25  
 STANDARD DEVIATION = 9.35617  
 NUMBER OF OBSERVATIONS = 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	87.5378	1.00000												*****										0
1	69.0171	0.78843												*****										0.182574
2	53.5328	0.61154												*****										0.273449
3	56.5819	0.64637												*****										0.315763
4	51.9773	0.59377												*****										0.357155
5	32.1162	0.36688												*****										0.388669
6	26.1013	0.29817												*****										0.400046
7	24.1731	0.27614												*****										0.407387
8	11.6782	0.13341												***										0.413358
9	-1.97825	-0.02260												**										0.415011
10	-7.87917	-0.09001												**										0.415052

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.64532												*****										
2	0.21583												****										
3	0.10623												**										
4	-0.42439												*****										
5	0.43180												*****										
6	-0.22277												****										
7	0.01671												****										
8	0.05386												*										
9	-0.09509												**										
10	0.07555												**										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.78843												*****										
2	-0.02663												*										
3	0.45587												*****										
4	-0.18277												****										
5	-0.32625												*****										
6	0.16977												****										
7	-0.23351												*****										
8	-0.02593												****										
9	-0.12438												****										
10	-0.21573												****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	70.00	6	0.000	0.788	0.612	0.646	0.594	0.367	0.298

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ARIMA PROCEDURE

NAME OF VARIABLE - B14  
 MEAN OF WORKING SERIES- 10.4167  
 STANDARD DEVIATION - 4.89027  
 NUMBER OF OBSERVATIONS- 30  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	23.9147	1.00000												*****										0
1	12.4696	0.52142												*****										0.182574
2	9.21081	0.38515												*****										0.226845
3	9.81681	0.41049												*****										0.247685
4	7.29857	0.30519												*****										0.26941
5	4.29523	0.17961												****										0.280698
6	3.49789	0.14627												***										0.284502
7	0.223269	0.00934												**										0.286998
8	-0.771685	-0.03227												*										0.287008
9	3.53575	0.14785												***										0.287139
10	-0.452426	-0.01892												**										0.289656

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.31910												*****										
2	-0.01593												**										
3	-0.10536											***											
4	-0.13177												*										
5	0.06135													*									
6	-0.00749													****									
7	0.05631													*****									
8	0.18166													**									
9	-0.28651																						
10	0.10810													**									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.52142												*****										
2	0.15557												****										
3	0.22490												****										
4	-0.00035																						
5	-0.07980												**										
6	-0.02123																						
7	-0.16427												***										
8	-0.02753												*										
9	0.27932													*****									
10	-0.15514												***										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	AUTOCORRELATIONS					
LAG	6	25.61	6	0.000	0.521	0.385	0.410	0.305	0.180	0.146

4 - ARIMA for Tojeira 1 and 2 sections with Gradstein's data



NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES = 21.6412  
 STANDARD DEVIATION = 11.8928  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
0	141.439	1.00000												*****	*****									
1	38.1513	0.26974										***												
2	-20.0548	-0.14179									*													
3	-4.20078	-0.02970										*												
4	4.50278	0.03184											*											
5	-5.85604	-0.04140												*										
6	-31.831	-0.22505													*									

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	-0.38177											*****												
2	0.26639												*****											
3	-0.12548													***										
4	0.10370														**									
5	-0.08331															**								
6	0.17286																***							

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.26974											*****												
2	-0.23138												*****											
3	0.09094													**										
4	-0.02217														*									
5	-0.04675															*								
6	-0.21617																****							

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	3.48	6	0.747	0.270	-0.142	-0.030	0.032	-0.041	-0.225

NAME OF VARIABLE = A3  
 MEAN OF WORKING SERIES = 6.40588  
 STANDARD DEVIATION = 5.68118  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	32.2758	1.00000												*****									
1	-1.97734	-0.06126									*												
2	2.95346	0.09151												**									
3	5.42415	0.16806												***									
4	-6.07136	-0.18811											****										
5	-8.05679	-0.24962											****										
6	-13.7782	-0.42689											*****										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.30357												*****										
2	-0.13706																						
3	-0.29889										***												
4	0.10453										*****												
5	0.32608												**										
6	0.33585												*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.06126										*												
2	0.08808												**										
3	0.18064												***										
4	-0.18192												****										
5	-0.32762												*****										
6	-0.55264												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	8.82	6	0.184	-0.061	0.092	0.168	-0.188	-0.250	-0.427





NAME OF VARIABLE = A6  
 MEAN OF WORKING SERIES = 2.41176  
 STANDARD DEVIATION = 1.31412  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
0	1.72692	1.00000												*****	*****										
1	0.763244	0.44197																							
2	-0.072473	-0.04197																							
3	-0.481824	-0.27901																							
4	-0.333424	-0.19307																							
5	0.0297171	0.01721																							
6	0.124865	0.07230																							

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	-0.43875												*****	*										
2	0.05191													**										
3	0.12306													*										
4	0.06143													**										
5	-0.09927													*										
6	0.04964																							

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.44197												*****											
2	-0.29491													*****										
3	-0.16966														***									
4	0.02709														*									
5	0.06389														*									
6	-0.06758														*									

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	6.86	6	0.334	0.442	-0.042	-0.279	-0.193	0.017	0.072

NAME OF VARIABLE = A7  
 MEAN OF WORKING SERIES = 2.72941  
 STANDARD DEVIATION = 2.1687  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	4.70325	1.00000												*****									
1	-0.0227498	-0.00484																					
2	-0.925361	-0.19675									****												
3	-1.59756	-0.33967								*****													
4	0.032357	0.00688											**										
5	-0.35676	-0.07585											*										
6	-0.129821	-0.02760																					

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.39905												*****										
2	0.42832												*****										
3	0.49050												*****										
4	0.19623												****										
5	0.21231												****										
6	0.15815												***										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.00484																						
2	-0.19678										****												
3	-0.35554									*****													
4	-0.07108										*												
5	-0.26194										*****												
6	-0.25259										*****												

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	3.67	6	0.721	-0.005	-0.197	-0.340	0.007	-0.076	-0.028

NAME OF VARIABLE = A8  
 MEAN OF WORKING SERIES= 7.09412  
 STANDARD DEVIATION = 4.53878  
 NUMBER OF OBSERVATIONS= 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	20.6006	1.00000												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
1	4.72761	0.22949												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
2	-0.795194	-0.03860									*			*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
3	-0.18575	-0.00902										*		*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
4	-5.54001	-0.26893											*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
5	-4.23105	-0.20539											*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
6	-3.65351	-0.17735											*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.24570												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
2	0.24780												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
3	-0.15264												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
4	0.26483												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
5	-0.00675												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
6	0.14920												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.22949												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
2	-0.09634												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
3	0.02430												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
4	-0.29442												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
5	-0.07585												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
6	-0.18121												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS									
6	4.95	6	0.550	0.229	-0.039	-0.009	-0.269	-0.205	-0.177				

NAME OF VARIABLE = A9  
 MEAN OF WORKING SERIES = 3.14706  
 STANDARD DEVIATION = 3.33363  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	11.1131	1.00000												*****									
1	-3.12923	-0.28158								*****				*									
2	0.370916	0.03338												*									
3	-3.017	-0.27148									*****			*									
4	0.334046	0.03006												*									
5	-0.891586	-0.08023										**		**									
6	1.18718	0.10683												**									

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.51174												*****										
2	0.38165												*****										
3	0.39483												*****										
4	0.23210												*****										
5	0.16276												***										
6	0.05390												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.28158												*****										
2	-0.04986												*										
3	-0.30014												*****										
4	-0.16052												***										
5	-0.18010												*****										
6	-0.08077												**										

AUTOCORRELATION CHECK FOR WHITE NOISE

TC	CHI	DF	PROB	AUTOCORRELATIONS				
LAG	SQUARE							
5	3.86	6	0.696	-0.282	0.033	-0.271	0.030	-0.080 0.107

NAME OF VARIABLE = A10  
 MEAN OF WORKING SERIES = 3.4  
 STANDARD DEVIATION = 3.29349  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	10.8471	1.00000												*****	*****	*****	*****	*****	*****	*****	*****	*****	*****
1	3.67412	0.33872												*****									
2	3.52882	0.32533												*****									
3	0.995294	0.09176												**									
4	-1.19529	-0.11020											**										
5	-1.03112	-0.09534											**										
6	-0.149412	-0.01377											**										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.12361											**											
2	-0.38734									*****		*											
3	-0.06015									*		*											
4	0.24724											*	*****										
5	0.07393											**	*										
6	-0.12352										**	*	*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.33872												*****										
2	0.23789												*****										
3	-0.08717										**												
4	-0.23041									*****													
5	-0.01366																						
6	0.16058																						

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	5.34	6	0.501	0.339	0.325	0.092	-0.110	-0.095	-0.014

NAME OF VARIABLE = A11  
 MEAN OF WORKING SERIES = 14.4235  
 STANDARD DEVIATION = 7.03241  
 NUMBER OF OBSERVATIONS = 17

AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
0	49.4547	1.00000												*****											
1	0.644812	0.01304																							
2	4.94326	0.09996												**											
3	-0.633246	-0.01280																							
4	-7.33985	-0.14842																							
5	-12.558	-0.25393																							
6	-16.3663	-0.33093																							

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.18872												****											
2	-0.02907												*											
3	-0.02585												*											
4	0.10697													**										
5	0.25151													*****										
6	0.27632													*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.01304																							
2	0.09980																							
3	-0.01544																							
4	-0.15969																							
5	-0.25656																							
6	-0.33594																							

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	5.72	6	0.455	0.013	0.100	-0.013	-0.148	-0.254	-0.331

NAME OF VARIABLE = A12  
 MEAN OF WORKING SERIES= 5.60588  
 STANDARD DEVIATION = 6.50325  
 NUMBER OF OBSERVATIONS= 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	42.2923	1.00000												*****									
1	-6.11872	-0.14468									***												
2	-8.38865	-0.19835									****												
3	-1.09246	-0.02583									*												
4	-8.35198	-0.19748									****												
5	7.22092	0.17074											***										
6	-4.75818	-0.11251										**											

' MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.28760												*****										
2	0.43729												*****										
3	0.20307												****										
4	0.30856												*****										
5	0.02857												*										
6	0.16185												***										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.14468											***											
2	-0.22397										****												
3	-0.10039										**												
4	-0.28798										*****												
5	0.05283												*										
6	-0.22173										****												

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	3.41	6	0.756	-0.145 -0.198 -0.026 -0.197 0.171 -0.113

NAME OF VARIABLE = A13  
 MEAN OF WORKING SERIES = 3.28824  
 STANDARD DEVIATION = 4.60395  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
0	21.1963	1.00000												*****											
1	6.96954	0.32881												*****											
2	2.68746	0.12679												***											
3	-1.21996	-0.05756										*													
4	-4.94429	-0.23326									*****														
5	-2.81963	-0.13302									***														
6	-0.211018	-0.00996																							

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	-0.23231									*****														
2	-0.12077									**														
3	-0.01969																							
4	0.19486												****											
5	0.02428																							
6	-0.06713										*													

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.32881												*****											
2	0.02094																							
3	-0.11807										**													
4	-0.20845									****														
5	0.01896																							
6	0.07759												**											

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	4.44	6	0.618	0.329 0.127 -0.058 -0.233 -0.133 -0.010

NAME OF VARIABLE = A14  
 MEAN OF WORKING SERIES = 12.3882  
 STANDARD DEVIATION = 10.2683  
 NUMBER OF OBSERVATIONS = 17  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
0	105.438	1.00000												*****											
1	21.6447	0.20528												****											
2	-4.39074	-0.04164										*													
3	-8.37037	-0.07939									**														
4	-0.807299	-0.00766										*													
5	-17.6811	-0.16769										***													
6	-28.1517	-0.26700										****													

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1			
1	-0.15208										***														
2	0.05062											*													
3	0.11229											**													
4	-0.04696											*													
5	0.09117											**													
6	0.19157											****													

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.20528												****											
2	-0.08747										**													
3	-0.05485										*													
4	0.01880																							
5	-0.18809										****													
6	-0.21330										****													

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	3.89	6	0.692	0.205	-0.042	-0.079	-0.008	-0.168	-0.267



NAME OF VARIABLE = B2  
 MEAN OF WORKING SERIES= 0.346154  
 STANDARD DEVIATION = 0.397328  
 NUMBER OF OBSERVATIONS= 13  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	0.15787	1.00000												*****									
1	-0.0376195	-0.23829											*****										
2	-0.0174875	-0.11077											**										
3	-0.00688711	-0.04359											*										
4	0.0223623	0.14165												***									
5	-0.0342513	-0.21696											****										
6	0.0293127	0.18568												****									

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.25588												*****										
2	0.22695												*****										
3	0.11221												**										
4	-0.03220											*											
5	0.11236												**										
6	-0.10546											**											

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.23829												*****										
2	-0.17764												*****										
3	-0.12810												***										
4	0.08263												**										
5	-0.19695												****										
6	0.12355												**										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS									
6	3.72	6	0.715	-0.238	-0.111	-0.044	0.142	-0.217	0.186				



NAME OF VARIABLE = B4  
 MEAN OF WORKING SERIES = 11.2769  
 STANDARD DEVIATION = 18.3383  
 NUMBER OF OBSERVATIONS = 13

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	336.294	1.00000												*****									
1	-43.2555	-0.12862									***												
2	-12.4749	-0.03710									*												
3	-46.4074	-0.13800									***												
4	97.8625	0.29100												*****									
5	-29.228	-0.08691									**												
6	9.77839	0.02908												*									

MARKS TWO STANDARD ERRORS

#### INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.06120											*											
2	0.06660											*											
3	0.08174											**											
4	-0.23384										*****	*											
5	0.03087											*											
6	-0.02726											*											

#### PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.12862											***											
2	-0.05454											*											
3	-0.15302											***											
4	0.25968														*****								
5	-0.03937											*											
6	0.02984											*											

#### AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS						
6	2.71	6	0.845	-0.129	-0.037	-0.138	0.291	-0.087	0.029	







*S*

NAME OF VARIABLE = B8  
 MEAN OF WORKING SERIES = 1.18462  
 STANDARD DEVIATION = 0.80369  
 NUMBER OF OBSERVATIONS = 13  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
0	0.645917	1.00000												*****										
1	-0.138302	-0.21412									****				***									
2	0.106532	0.16493										***												
3	-0.082244	-0.12733										****												
4	-0.14741	-0.22822										****												
5	-0.0596177	-0.09230										**												
6	-0.185435	-0.28709										*****												

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.38862												*****										
2	0.13082													***									
3	0.19806													***									
4	0.27880													*****									
5	0.27946													*****									
6	0.24361													*****									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.21412										****												
2	0.12481												**										
3	-0.07420										*												
4	-0.30518									*****													
5	-0.19340									****													
6	-0.34483									*****													

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	5.18	6	0.522	-0.214	0.165	-0.127	-0.228	-0.092	-0.287

*J*



NAME OF VARIABLE = B10  
 MEAN OF WORKING SERIES = 1.57692  
 STANDARD DEVIATION = 1.6821  
 NUMBER OF OBSERVATIONS = 13  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
0	2.82947	1.00000												*****										
1	0.8299	0.29331												*****										
2	0.296013	0.10462												**										
3	0.0594037	0.02099												*										
4	0.142321	0.05030																						
5	-0.462039	-0.16330											***											
6	-0.566163	-0.20010											***											

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.23801											*****											
2	-0.03799											*											
3	0.05151												*										
4	-0.14489											***											
5	0.12467												**										
6	0.10326												**										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.29331												*****										
2	0.02034																						
3	-0.01645																						
4	0.05087												*										
5	-0.20854											****											
6	-0.11708											**											

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS									
6	3.42	6	0.754	0.293	0.105	0.021	0.050	-0.163	-0.200				



NAME OF VARIABLE = B12  
 MEAN OF WORKING SERIES = 1.36923  
 STANDARD DEVIATION = 1.08087  
 NUMBER OF OBSERVATIONS = 13  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
0	1.16828	1.00000												*****											
1	0.211821	0.18131												****											
2	0.174292	0.14919												***											
3	-0.467497	-0.40016																							
4	-0.156268	-0.13376																							
5	-0.238885	-0.20447																							
6	0.0384388	0.03290																							

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	-0.21255												****											
2	-0.20411												****											
3	0.42260																							
4	-0.04496												*											
5	-0.00511																							
6	0.07887																							

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
1	0.18131												****											
2	0.12027												**											
3	-0.46767																							
4	0.01114																							
5	-0.03989												*											
6	-0.10672												**											

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	5.49	6	0.483	0.181	0.149	-0.400	-0.134	-0.204	0.033

NAME OF VARIABLE = B13  
 MEAN OF WORKING SERIES = 3.16923  
 STANDARD DEVIATION = 3.83734  
 NUMBER OF OBSERVATIONS = 13  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1		
0	14.7252	1.00000												*****											
1	2.48354	0.16866												***											
2	-5.25287	-0.35673									*****														
3	-2.11158	-0.14340									***														
4	0.80965	0.05498										*													
5	-1.19457	-0.08112										**													
6	-3.91594	-0.26593										*****													

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.14213										***												
2	0.39361												*****										
3	0.06957												*										
4	0.17761												****										
5	0.04167												*										
6	0.20250												****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.16866												***										
2	-0.39645									*****													
3	0.00914																						
4	-0.06651										*												
5	-0.16575										***												
6	-0.26785										*****												

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS									
6	5.32	6	0.504	0.169	-0.357	-0.143	0.055	-0.081	-0.266				

NAME OF VARIABLE = B14  
 MEAN OF WORKING SERIES = 19.7154  
 STANDARD DEVIATION = 18.2136  
 NUMBER OF OBSERVATIONS = 13  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
0	331.734	1.00000												*****									
1	-36.3239	-0.10950									**												
2	-31.2764	-0.09428									**												
3	-45.2304	-0.13635									***												
4	131.488	0.39637												*****									
5	-114.864	-0.34625									*****												
6	-23.085	-0.06959									*												

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.13984										***												
2	0.08227												**										
3	0.16055												***										
4	-0.26739										*****												
5	0.27948												*****										
6	0.03312												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.10950										**												
2	-0.10756										**												
3	-0.16338										***												
4	0.36563												*****										
5	-0.36007										*****												
6	-0.04222										*												

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI				AUTOCORRELATIONS							
LAG	SQUARE	DF	PROB									
6	7.18	6	0.305	-0.109	-0.094	-0.136	0.396	-0.346	-0.070			

5 - ARIMA for Tojeira 1 section with enlarged data

SAS

12:55 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE - A1  
 MEAN OF WORKING SERIES - 7.31951  
 STANDARD DEVIATION - 6.05049  
 NUMBER OF OBSERVATIONS - 41

LAG COVARIANCE CORRELATION		AUTOCORRELATIONS																					STD
		-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
0	36.6084	1.00000	*****																				0
1	8.54141	0.23332		*****																			0.156174
2	1.10711	0.03024			*****																		0.164456
3	-4.74188	-0.12953				*****																0.164591	
4	-2.14367	-0.05856					*****														0.167059		
5	2.84364	0.07768						*****												0.167559			
6	3.96367	0.10827							*****										0.168435				
7	3.71424	0.10146								*****								0.170124					
8	-4.37248	-0.11944									*****						0.171594						
9	0.815104	0.02227										*****				0.173609							
10	-4.14431	-0.11321											*****		0.173679								

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
1	-0.27292		*****																			
2	-0.00133			*****																		
3	0.07235				*****																	
4	0.03253					*****																
5	-0.05115						*****															
6	-0.04249							*****														
7	-0.08777								*****													
8	0.18613									*****												
9	-0.15838										*****											
10	0.11190											*****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1
1	0.23332		*****																			
2	-0.02559			*****																		
3	-0.13842				*****																	
4	0.00437					*****																
5	0.10324						*****															
6	0.05408							*****														
7	0.05345								*****													
8	-0.15057									*****												
9	0.11702										*****											
10	-0.13093											*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PRCB	AUTOCORRELATIONS					
6	4.27	6	0.640	0.233	0.030	-0.130	-0.059	0.078	0.108

12:55 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE = A2  
 MEAN OF WORKING SERIES= 2.20244  
 STANDARD DEVIATION = 4.03343  
 NUMBER OF OBSERVATIONS= 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	16.2685	1.00000												*****											0	
1	8.39487	0.51602													*****											0.156174
2	4.3428	0.26694													*****											0.193337
3	5.46661	0.33602													*****											0.202127
4	5.0157	0.30831													*****											0.215321
5	1.59018	0.09775													**											0.225832
6	0.546632	0.03360													*											0.226861
7	0.171189	0.01052													*											0.226983
8	0.0428721	0.00264													*											0.226995
9	-0.487336	-0.02996													*											0.226995
10	-0.418818	-0.02574													*											0.227092

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.42039												*****										
2	-0.14653													***									
3	-0.13343													***									
4	-0.16956													***									
5	-0.12696													*									
6	-0.05874													*									
7	-0.10302													**									
8	-0.03658													*									
9	-0.02154													*									
10	-0.01973													*									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.51602												*****										
2	0.00091													*****									
3	0.26976													*									
4	0.04826													*									
5	-0.15205													***									
6	-0.02644													**									
7	-0.08927													*									
8	0.03581													*									
9	-0.00216													*									
10	0.02737													*									

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	25.25	6	0.000	0.267 0.336 0.308 0.098 0.034

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ARIMA PROCEDURE

NAME OF VARIABLE = A3  
 MEAN OF WORKING SERIES = 23.7561  
 STANDARD DEVIATION = 12.1219  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD	
0	146.941	1.00000												*****											0.156174
1	66.4589	0.45228												*****											0.185388
2	77.2056	0.52542												*****											0.218713
3	41.5462	0.28274												*****											0.227453
4	31.3431	0.21336												*****											0.232281
5	22.3485	0.15209												*****											0.234697
6	7.02954	0.04784												*****											0.234935
7	-5.86578	-0.03992												*****											0.2351
8	5.27999	0.03593												*****											0.235234
9	-7.73063	-0.05261												*****											0.235521
10	6.02891	0.04103												*****											

MARKS TWO STANDARD ERRORS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.16148													*****									
2	-0.39178													*****									
3	0.10091													*****									
4	0.04509													*****									
5	-0.16663													*****									
6	0.07035													*****									
7	0.14896													*****									
8	-0.09290													*****									
9	0.01554													*****									
10	-0.01829													*****									

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.45228													*****									
2	0.40337													*****									
3	-0.06228													*****									
4	-0.09450													*****									
5	0.03227													*****									
6	-0.06248													*****									
7	-0.13608													*****									
8	0.14618													*****									
9	-0.00600													*****									
10	0.02536													*****									

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI																							
LAG	SQUARE	DF	PROB	0.452	0.525	0.283	0.213	0.152	0.048															
6	28.62	6	0.000																					

AUTOCORRELATIONS

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ARIMA PROCEDURE

NAME OF VARIABLE = A4  
 MEAN OF WORKING SERIES = 4.46341  
 STANDARD DEVIATION = 5.04597  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	25.4618	1.00000												*****										0.156174
1	6.39527	0.24999												****										0.165647
2	-1.38245	-0.05430												**										0.16608
3	3.2394	0.12723												***										0.168441
4	2.26237	0.08885												**										0.16958
5	-3.02327	-0.11874												***										0.171596
6	3.87057	0.15201												****										0.174849
7	6.58025	0.25844												**										0.18393
8	-2.63504	-0.10349												***										0.185345
9	-3.73567	-0.14672												***										0.188156
10	-4.281	-0.16813												***										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.30566												*****										
2	0.10194												**										
3	-0.03355												*										
4	-0.17787												****										
5	0.19071												***										
6	-0.10213												**										
7	-0.21402												****										
8	0.17069												***										
9	-0.08118												**										
10	0.20016												****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.24999												*****										
2	-0.12458												**										
3	0.18813												****										
4	-0.00737												**										
5	-0.12468												****										
6	0.24357												**										
7	0.11698												****										
8	-0.18142												**										
9	-0.05843												****										
10	-0.27370												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

AUTOCORRELATIONS  
 TO CHI SQUARE DF PROB 0.250 -0.054 0.127 0.089 -0.119 0.152 12:55 MONDAY, OCTOBER 5, 1987  
 LAG 6 5.87 6 0.438 SAS

ARIMA PROCEDURE

NAME OF VARIABLE - A5  
 MEAN OF WORKING SERIES - 2.52439  
 STANDARD DEVIATION - 3.55574  
 NUMBER OF OBSERVATIONS - 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	12.6433	1.00000												*****											0	
1	2.24441	0.17752												****											0.156174	
2	-1.30087	-0.10289												****											0.16102	
3	-3.11962	-0.24674												****											0.162616	
4	-2.44439	-0.19333												****											0.171504	
5	-0.848978	-0.06715												****											0.17674	
6	0.133595	0.01057												****											0.177361	
7	-0.184646	-0.01460												****											0.177376	
8	-1.29605	-0.10251												****											0.177406	
9	-0.5111	-0.04042												****											0.178845	
10	-0.868027	-0.07024												****											0.179067	

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.08220												****										
2	0.26103												****										
3	0.29717												****										
4	0.24542												****										
5	0.19106												****										
6	0.14600												****										
7	0.13330												****										
8	0.16722												****										
9	0.03215												****										
10	0.12786												****										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.17752												****										
2	-0.13878												****										
3	-0.21194												****										
4	-0.13554												****										
5	-0.06988												****										
6	-0.05981												****										
7	-0.10565												****										
8	-0.17006												****										
9	-0.06913												****										
10	-0.16106												****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB *	AUTOCORRELATIONS					
6	6.70	6	0.350	0.178	-0.103	-0.247	-0.193	-0.067	0.011

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ARIMA PROCEDURE

NAME OF VARIABLE = A6  
 MEAN OF WORKING SERIES = 11.5854  
 STANDARD DEVIATION = 6.37172  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	40.5988	1.00000												*****											0	
1	12.1566	0.29943										*													0.156174	
2	-1.2492	-0.03077									*	*													0.169599	
3	-10.8471	-0.26718									*	*	*												0.169735	
4	-10.1845	-0.25086									*	*	*	*											0.1797	
5	-7.2753	-0.17925									*	*	*	*	*										0.188048	
6	-1.41672	-0.03490									*	*	*	*	*	*									0.19217	
7	2.3766	0.05854									*	*	*	*	*	*	*								0.192325	
8	2.82059	0.06947									*	*	*	*	*	*	*	*								0.192759
9	1.55476	0.03830									*	*	*	*	*	*	*	*	*							0.193368
10	-5.70755	-0.14058									*	*	*	*	*	*	*	*	*	*						0.193553

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.16618										***												
2	0.08851										*	*											
3	0.21336										*	*	*										
4	0.08943										*	*	*	*									
5	0.14137										*	*	*	*	*								
6	0.05964										*	*	*	*	*	*							
7	0.07467										*	*	*	*	*	*	*						
8	-0.06213										*	*	*	*	*	*	*	*					
9	-0.05682										*	*	*	*	*	*	*	*	*				
10	0.16444										*	*	*	*	*	*	*	*	*	*			

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.29943										***												
2	-0.13229										*	*											
3	-0.24277										*	*	*										
4	-0.11607										*	*	*	*									
5	-0.11543										*	*	*	*	*								
6	-0.03955										*	*	*	*	*	*							
7	-0.02149										*	*	*	*	*	*	*						
8	-0.04282										*	*	*	*	*	*	*	*					
9	-0.03244										*	*	*	*	*	*	*	*	*				
10	-0.19869										*	*	*	*	*	*	*	*	*	*			

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	11.94	6	0.063	0.299 -0.031 -0.267 -0.251 -0.179 -0.035

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ARIMA PROCEDURE

NAME OF VARIABLE = A7  
 MEAN OF WORKING SERIES = 4.08293

STANDARD DEVIATION = 2.714  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD		
0	7.36581	1.00000												*****											0	
1	1.09548	0.14873												***											0.156174	
2	0.21186	0.02876												***											0.159591	
3	1.43856	0.19530												****											0.159717	
4	-0.336459	-0.04568																							0.165439	
5	-0.823165	-0.11175									**														0.165747	
6	0.0338657	0.00460										*													0.167575	
7	0.231467	0.03142																							0.167578	
8	-1.63057	-0.22816									*****														0.167721	
9	0.580703	0.07884										**													0.175128	
10	-0.635859	-0.08633										**													0.175991	

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.31871											*****											
2	0.15854											*****	***										
3	-0.23596											*****	***										
4	0.15216												***										
5	-0.04968												*										
6	0.13650												***										
7	-0.19177											****											
8	0.24107												*****										
9	-0.18119											****											
10	0.14007												***										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.14873												***										
2	0.00679													****									
3	-0.19435														****								
4	-0.10899															****							
5	-0.09594																****						
6	-0.00186																	****					
7	0.06613																		****				
8	-0.22337																			****			
9	0.15267																				****		
10	-0.17870																					****	

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	3.49	6	0.745	0.149 0.029 0.195 -0.046 -0.112 0.005

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ARIMA PROCEDURE

NAME OF VARIABLE = A8  
 MEAN OF WORKING SERIES = 2.15122  
 STANDARD DEVIATION = 1.26436  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	1.5986	1.00000												*****										0.156174
1	0.290697	0.18185												****										0.161255
2	0.301811	0.18880												****										0.166559
3	-0.431037	-0.26963												***										0.176886
4	-0.23768	-0.14868																						0.179908
5	0.0227675	0.01424																						0.179935
6	0.225232	0.14089																						0.182606
7	0.30029	0.18785																						0.18726
8	0.015086	0.00944																						0.187272
9	-0.280487	-0.17546																						0.191239
10	-0.236024	-0.14764																						

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.14426												****										
2	-0.36151												****										
3	0.23642												**										
4	0.08860												*										
5	-0.05723																						
6	0.00630												**										
7	-0.11918																						
8	0.01418																						
9	0.13225																						
10	-0.01834																						

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.18185												****										
2	0.16106												*****										
3	-0.34796												**										
4	-0.08406													****									
5	0.22041													**									
6	0.08156																						
7	0.01606																						
8	-0.06229																						
9	-0.16564																						
10	0.02412																						

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI																							
LAG	SQUARE	DF	PROB	0.182	0.189	-0.270	-0.149	0.014	0.141															
6	8.50	6	0.203																					

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ARIMA PROCEDURE

NAME OF VARIABLE - A9  
 MEAN OF WORKING SERIES - 3.37317  
 STANDARD DEVIATION - 5.856  
 NUMBER OF OBSERVATIONS - 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	34.2927	1.00000												*****										0.156174
1	0.39959	0.01165												**										0.156195
2	2.69638	0.07863												****										0.157157
3	7.08272	0.20654												**										0.163644
4	-2.64996	-0.07727												*										0.164531
5	1.88719	0.05503												*										0.16498
6	-1.36239	-0.03973												***										0.165213
7	-4.46498	-0.13020												**										0.167697
8	2.77023	0.08078												*										0.168643
9	0.398515	0.01162												*										0.168663
10	-1.05833	-0.03086												*										

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.00220												**										
2	-0.08767											****											
3	-0.20863											*											
4	0.02158											*											
5	-0.01919											*											
6	0.10036											*	**										
7	0.10788											*	**										
8	-0.06717											*	*										
9	-0.05697											*	*										
10	0.00400											*	*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.01165												**										
2	0.07850												****										
3	-0.20608												*										
4	-0.08886												**										
5	0.02515												*										
6	-0.07404												**										
7	-0.10716												**										
8	0.07403												*										
9	0.06176												*										
10	-0.00434												*										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	AUTOCORRELATIONS	
LAG	SQUARE	DF	PROB
6	2.78	6	0.836
			0.012
			0.079
			0.207
			-0.077
			0.055
			-0.040

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ARIMA PROCEDURE

NAME OF VARIABLE - A10  
 MEAN OF WORKING SERIES- 2.39268  
 STANDARD DEVIATION - 2.90906  
 NUMBER OF OBSERVATIONS- 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	8.46263	1.00000												*****										0
1	1.52864	0.18063												****										0.156174
2	4.52344	0.53452												*****										0.161189
3	3.91379	0.46248												*****										0.199797
4	1.8488	0.21847												*****										0.224394
5	1.94111	0.22937												*****										0.229523
6	2.58448	0.30540												*****										0.235047
7	0.096249	0.01137												****										0.244534
8	1.25868	0.14873												***										0.244547
9	1.46996	0.17370												**										0.246743
10	-0.903666	-0.10678												*										0.249708

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.19168												****										
2	-0.53654												*****										
3	-0.41434												*****										
4	0.06252												*										
5	0.21557												****										
6	0.07009												*										
7	0.01076												*										
8	-0.07735												**										
9	-0.06292												*										
10	0.06077												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.18063												****										
2	0.51882												*****										
3	0.45589												*****										
4	-0.09076												**										
5	-0.37582												*****										
6	0.06347												*										
7	0.04744												*										
8	-0.04929												*										
9	0.15305												***										
10	0.12346												**										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO	CHI	SQUARE	DF	PROB	LAG	AUTOCORRELATIONS
6	33.83	6	0.000	0.181	0.535	0.462 0.218 0.229 0.305

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ARIMA PROCEDURE

NAME OF VARIABLE = All  
 MEAN OF WORKING SERIES = 7.83171  
 STANDARD DEVIATION = 6.60799  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0		1.00000												*****										0.156174
1	43.6656	0.36446												*****										0.175698
2	15.9143	0.34056												*****										0.191122
3	14.8708	0.13720												***										0.193509
4	5.99078	0.01247																						0.193528
5	-0.544578	-0.01595																						0.193556
6	-0.696286	-0.02234																						0.193623
7	-0.975483	-0.02880																						0.199563
8	-9.55388	-0.22726										****												0.205778
9	-9.92328	-0.02549										****												0.205855
10	-1.11285	-0.24087										****												

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.25272												*****										
2	-0.34140												*****										
3	0.15694													***									
4	0.16400													***									
5	-0.12574																						
6	-0.17890																						
7	0.24895																						
8	0.11475																						
9	-0.26906																						
10	0.15120																						

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.36446												*****										
2	0.33955												*****										
3	-0.05444												**										
4	-0.11561												**										
5	-0.01146																						
6	0.02921																						
7	-0.24283												*****										
8	-0.14338												***										
9	0.26400												*****										
10	-0.23326												*****										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS					
6	12.02	6	0.062	0.364	0.341	0.137	0.012	-0.016	-0.022

ARIMA PROCEDURE

NAME OF VARIABLE - A12  
 MEAN OF WORKING SERIES - 3.65122  
 STANDARD DEVIATION - 2.69327  
 NUMBER OF OBSERVATIONS - 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD
0	7.25372	1.00000																						0
1	3.51547	0.48434												*****										0.156174
2	1.45065	0.19999												*****										0.189297
3	-0.117878	-0.01625												*****										0.194382
4	-1.0721	-0.14780												*****										0.194415
5	-1.31381	-0.18112												*****										0.197137
6	-2.27622	-0.31380												*****										0.201155
7	-2.19445	-0.30253												*****										0.21276
8	-1.39861	-0.19281												*****										0.223005
9	-0.705341	-0.09724												*****										0.227035
10	-0.655185	-0.09032												*****										0.228048

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.32680												*****										
2	-0.03779												*	**									
3	0.08899												*	**	*								
4	0.14713												*	**	*	*							
5	-0.13618												*	**	*	*	*						
6	0.14404												*	**	*	*	*	*					
7	0.10209												*	**	*	*	*	*	*				
8	-0.00514												*	**	*	*	*	*	*	*			
9	-0.07150												*	**	*	*	*	*	*	*	*		
10	0.12032												*	**	*	*	*	*	*	*	*	*	

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.48434												*****										
2	-0.04520												*	**									
3	-0.12515												*	**	*								
4	-0.10996												*	**	*	*							
5	-0.05595												*	**	*	*	*						
6	-0.24555												*	**	*	*	*	*					
7	-0.09228												*	**	*	*	*	*	*				
8	-0.00202												*	**	*	*	*	*	*	*			
9	-0.04510												*	**	*	*	*	*	*	*	*		
10	-0.15738												*	**	*	*	*	*	*	*	*	*	

AUTOCORRELATION CHECK FOR WHITE NOISE

TO CHI SQUARE DF PROB AUTOCORRELATIONS  
 LAG 6 19.77 6 0.003 0.484 0.200 -0.016 -0.148 -0.181 -0.314  
 SAS 12:55 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE = A13  
 MEAN OF WORKING SERIES = 2.15366  
 STANDARD DEVIATION = 5.07668  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD	
0	25.7727	1.00000												*****											0
1	10.1469	0.39371												*****											0.156174
2	3.06424	0.11889												**											0.17875
3	5.17424	0.20076												****											0.180668
4	3.83261	0.14871												****											0.18603
5	1.1649	0.04520												*											0.188907
6	-0.758916	-0.02945												*											0.189171
7	-1.31927	-0.05119												*											0.189283
8	0.844001	0.03275												*											0.18962
9	0.934288	0.03625												*											0.189758
10	-1.02274	-0.03968												*											0.189927

MARKS TWO STANDARD ERRORS

INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.38930												*****										
2	0.15081												***										
3	-0.17483												*										
4	-0.01312												*										
5	0.00752												*										
6	0.01971												*										
7	0.06654												*										
8	-0.03965												*										
9	-0.03865												*										
10	0.03133												*										

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.39371												*****										
2	-0.04273												*										
3	0.20011												****										
4	0.00079												*										
5	-0.01989												*										
6	-0.07818												**										
7	-0.04449												*										
8	0.07790												**										
9	0.01562												*										
10	-0.03899												*										

AUTOCORRELATION CHECK FOR WHITE NOISE

TO LAG	CHI SQUARE	DF	PROB	AUTOCORRELATIONS
6	10.54	6	0.104	0.394 0.119 0.201 0.149 0.045 -0.029

SAS 12:55 MONDAY, OCTOBER 5, 1987

ARIMA PROCEDURE

NAME OF VARIABLE = A14  
 MEAN OF WORKING SERIES = 15.7829  
 STANDARD DEVIATION = 8.64627  
 NUMBER OF OBSERVATIONS = 41  
 AUTOCORRELATIONS

LAG	COVARIANCE	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	STD	
0	74.758	1.00000												*****											0
1	6.26782	0.08384												**											0.156174
2	0.896922	0.01200												***											0.157268
3	10.5644	0.14131												*****											0.15729
4	20.4752	0.27389												*											0.160357
5	-3.53611	-0.04730												*											0.171387
6	-1.66831	-0.02232												*****											0.171705
7	17.0853	0.22854												*											0.171776
8	-4.33257	-0.05795												*											0.179039
9	0.30896	0.00413												*											0.179496
10	2.06242	0.02759												*											0.179498

MARKS TWO STANDARD ERRORS  
 INVERSE AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	-0.14297												***										
2	-0.01420												*										
3	0.02498												*****										
4	-0.28409												*										
5	0.08802												**										
6	0.05981												****										
7	-0.17890												*										
8	0.14685												***										
9	-0.05857												*										
10	0.00309																						

PARTIAL AUTOCORRELATIONS

LAG	CORRELATION	-1	9	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	1	
1	0.08384												**										
2	0.00500												*										
3	0.14089												***										
4	0.25735												*****										
5	-0.09059												*										
6	-0.03660												**										
7	0.18226												****										
8	-0.15832												*										
9	0.06816												***										
10	-0.00364												*										

AUTOCORRELATION CHECK FOR WHITE NOISE

AUTOCORRELATIONS  
 TO CHI SQUARE DF PROB SAS 0.012 0.141 0.274 -0.047 -0.022  
 LAG 6 4.95 6 0.550 0.084 SAS  
 12:55 MONDAY, OCTOBER 5, 1987

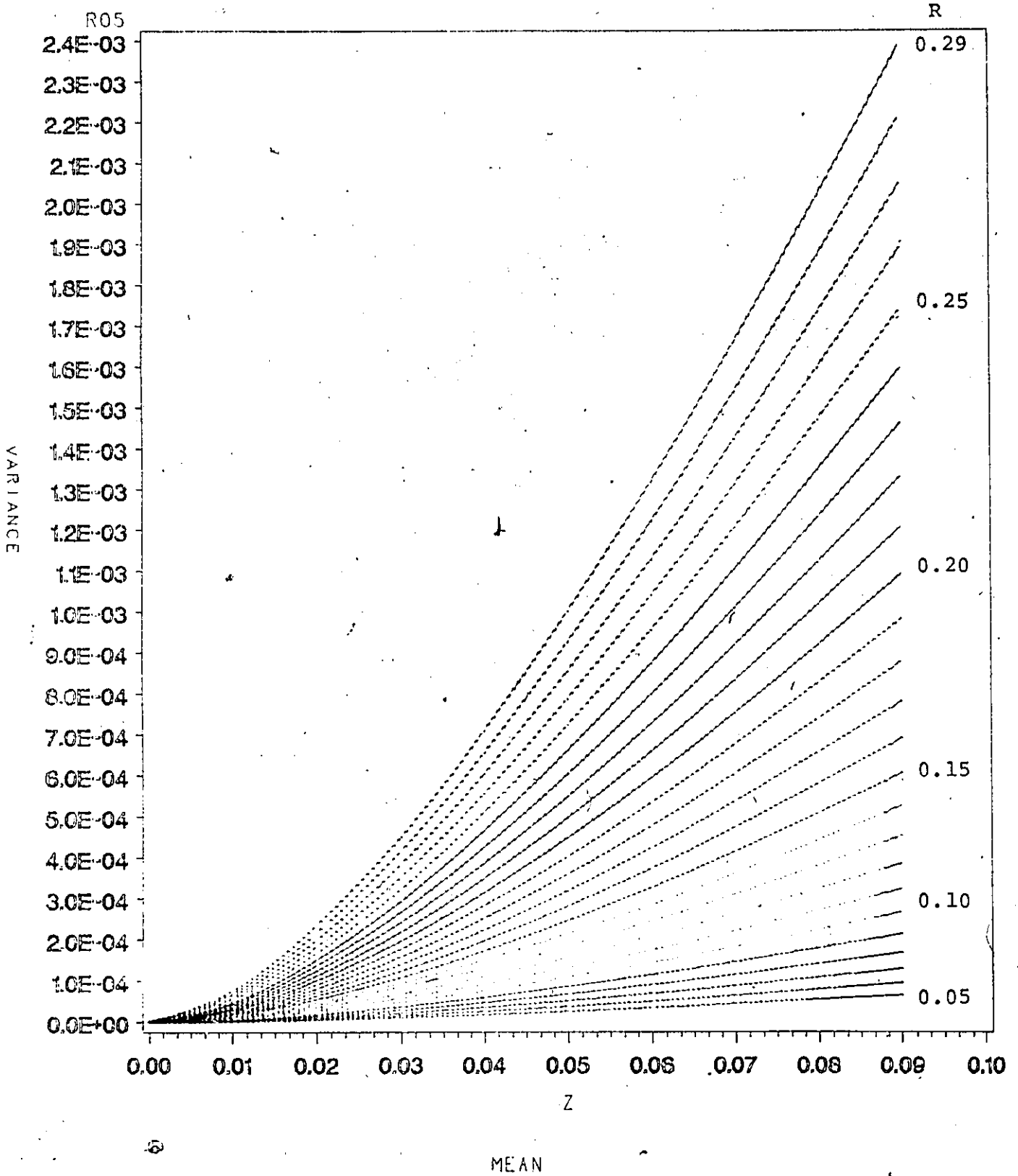
APPENDIX II: Relationships between mean and variance for  
various R values (GRAPHS)

- 1 - Relationships between mean and variance
- 2 - Relationships between mean and variance/(mean(1.0 - mean))
- 3 - Relationships between logarithmic mean and variance/(mean(1.0 - mean))

1 - Relationships between mean and variance

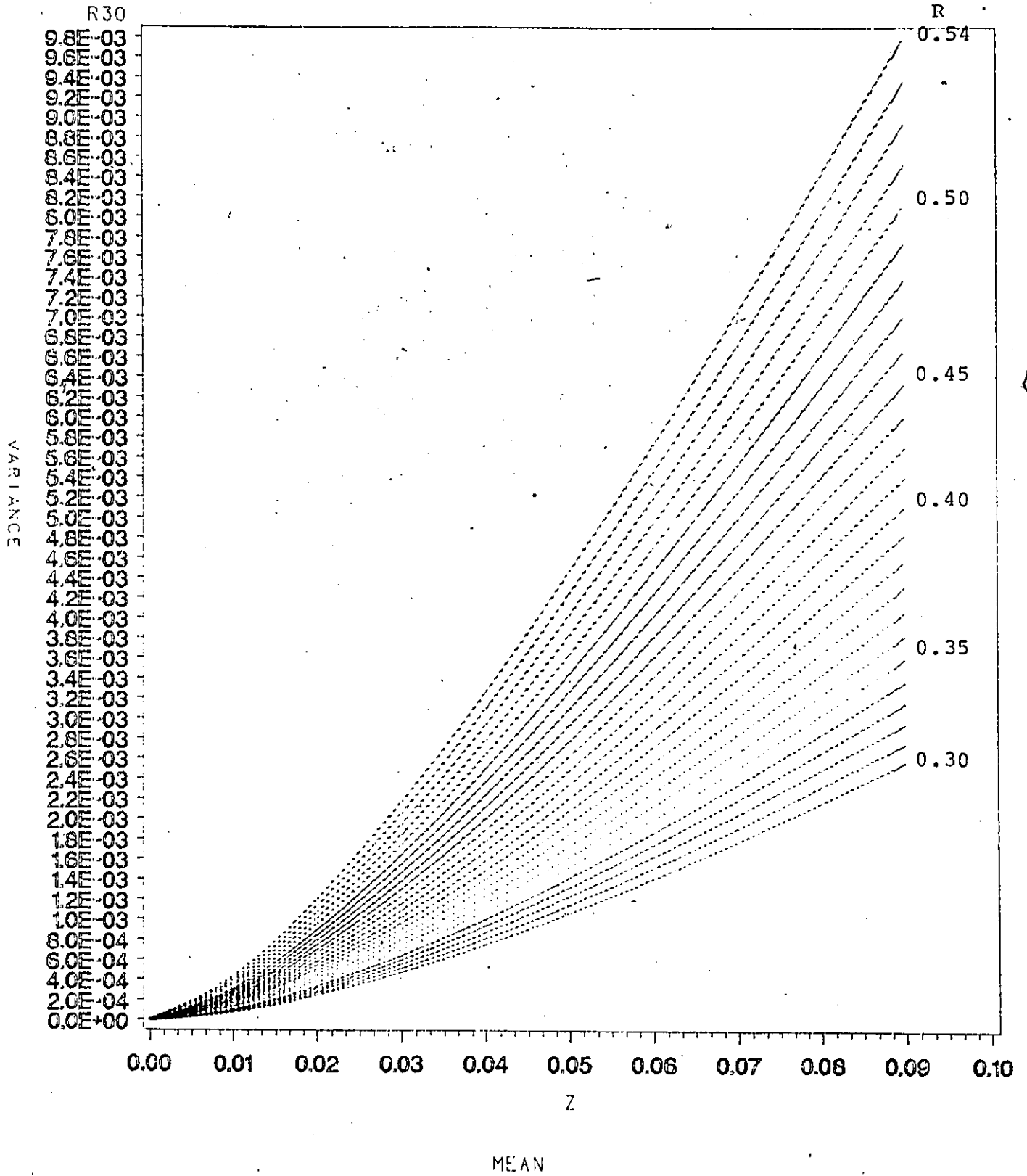
WITH SPLINE FITTING

344



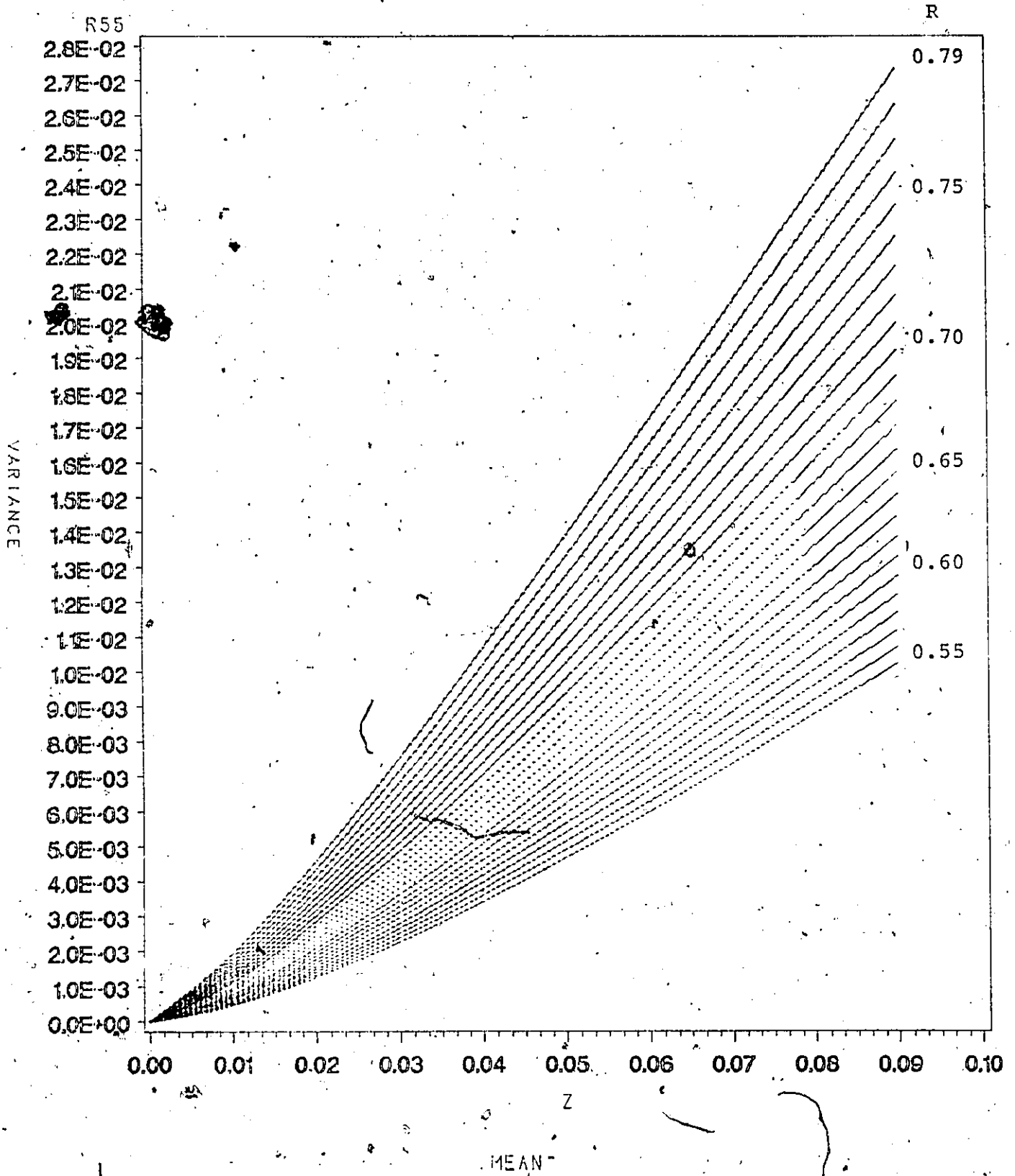
# RELATIONSHIP BETWEEN

WITH SPLINE FITTING

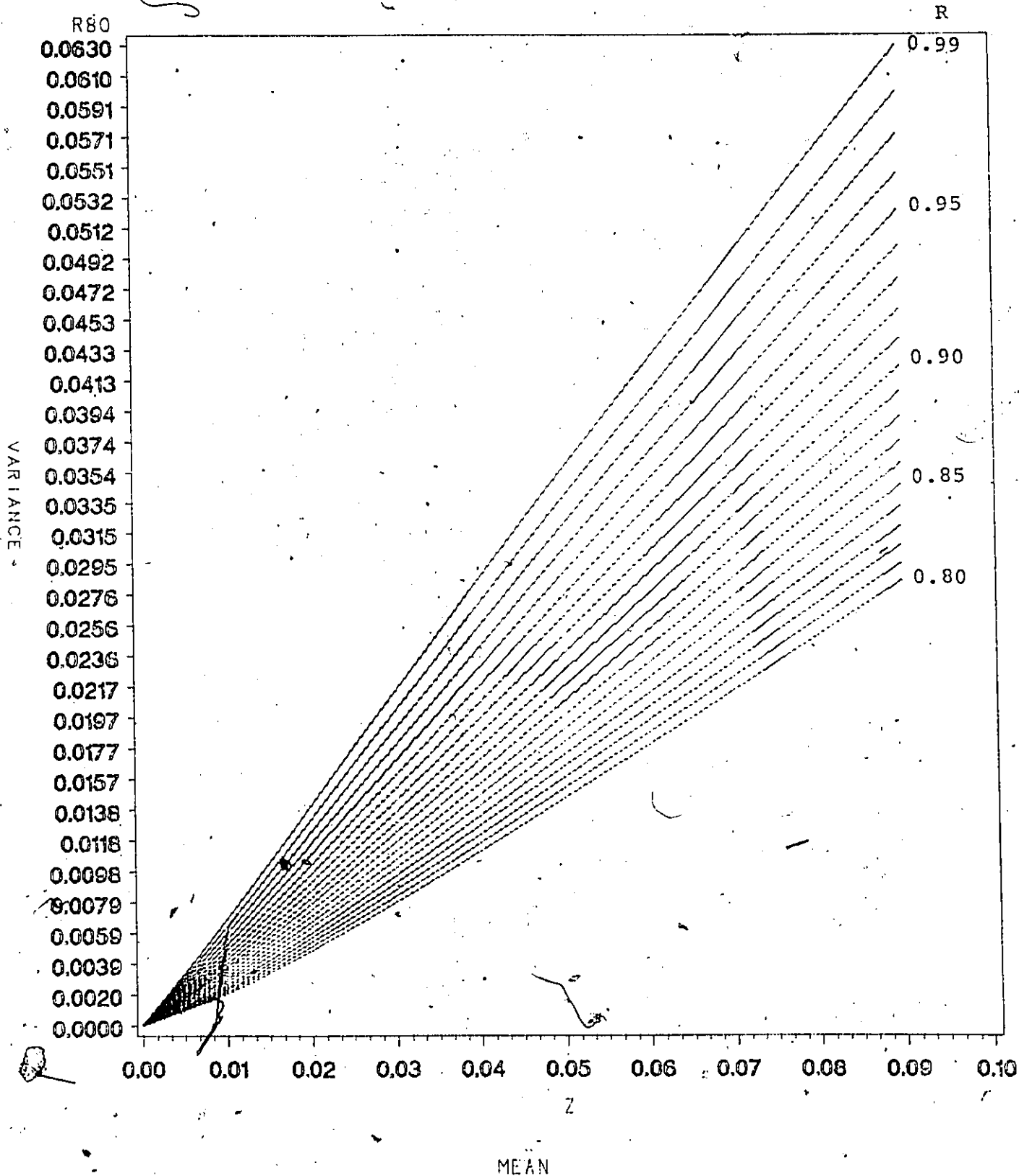


WITH SPLINE FITTING

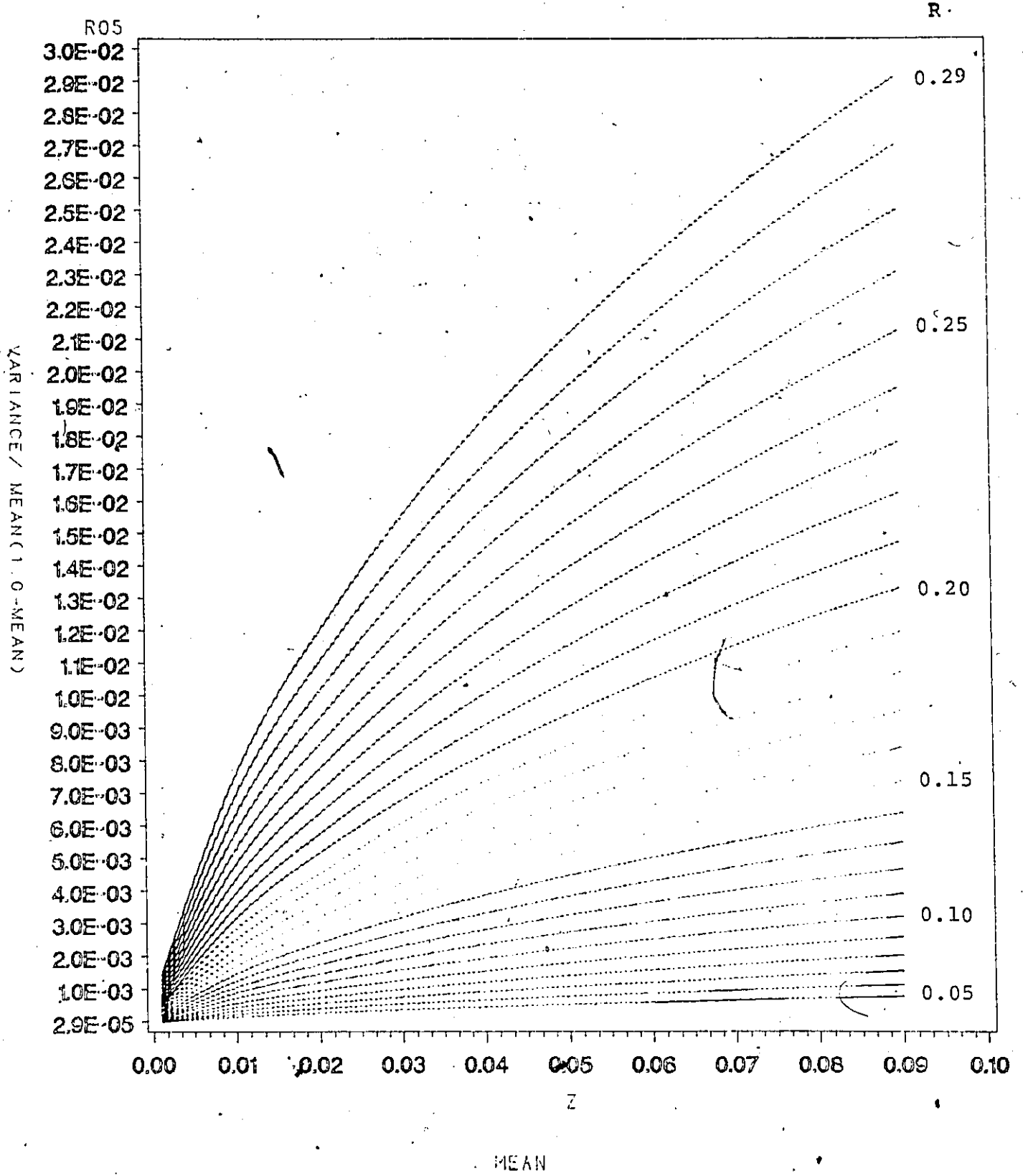
346



WITH SPLINE FITTING

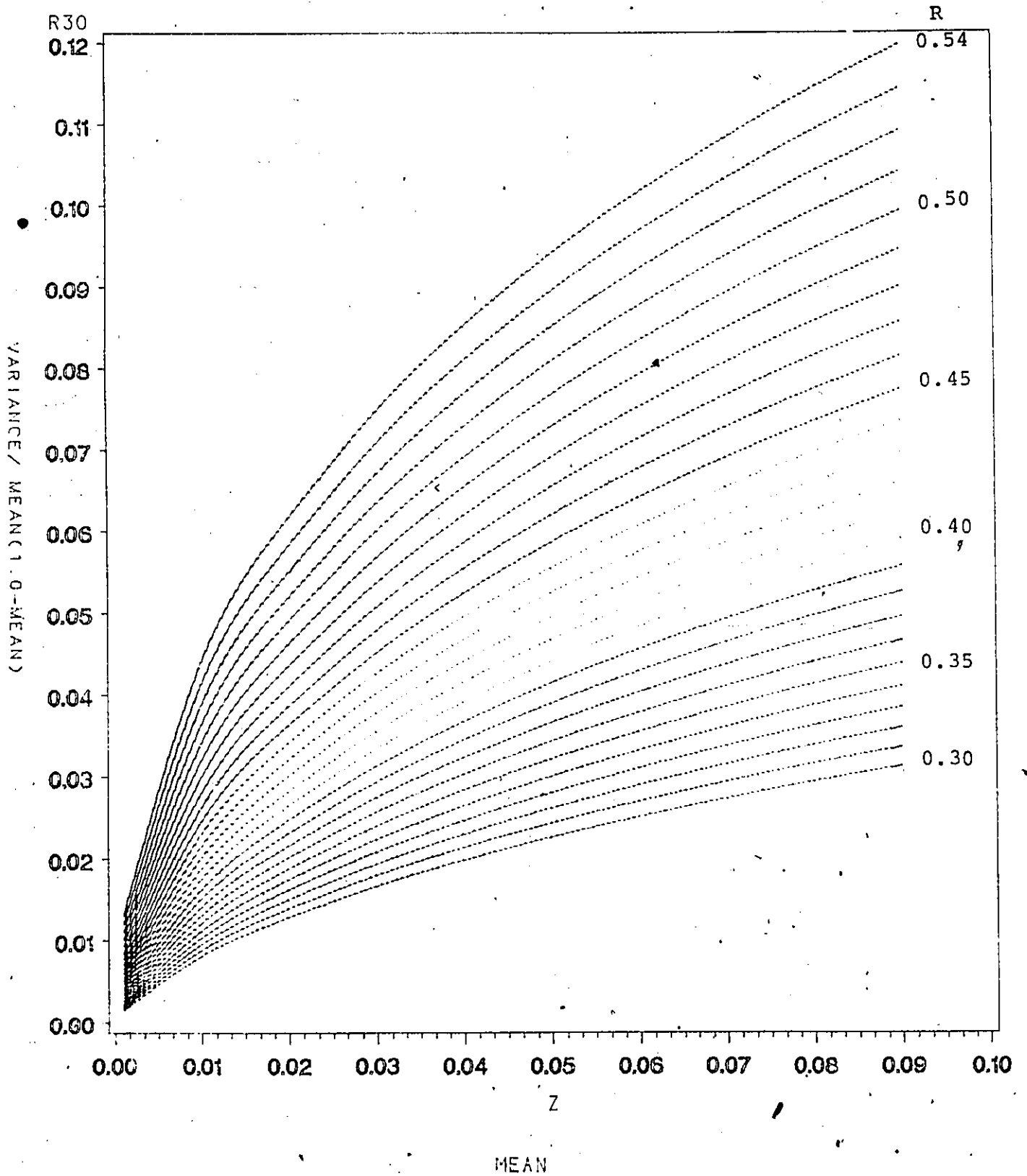


2 - Relationships between mean and variance/(mean(1.0 -  
mean ))



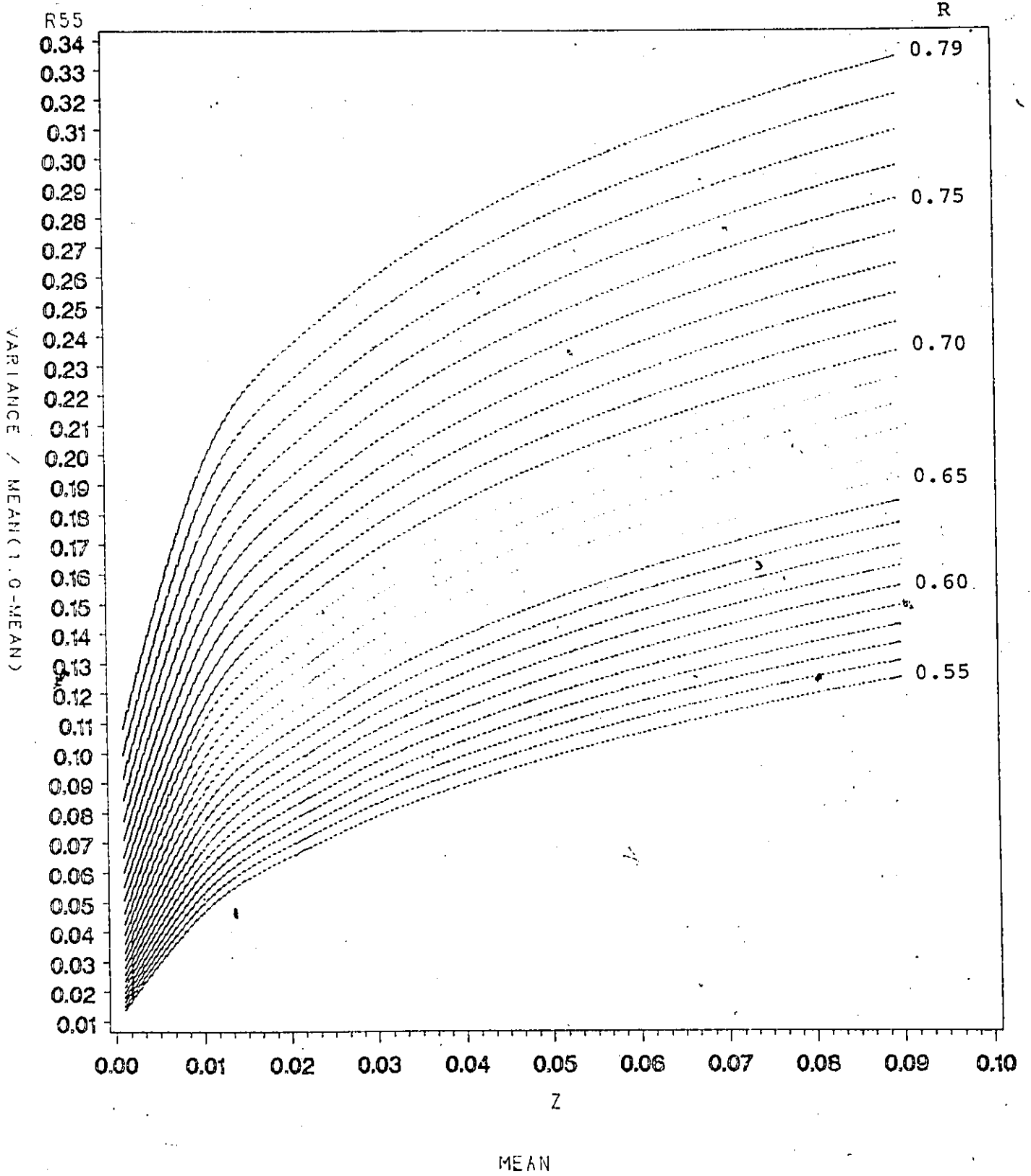
WITH SPLINE FITTING

350

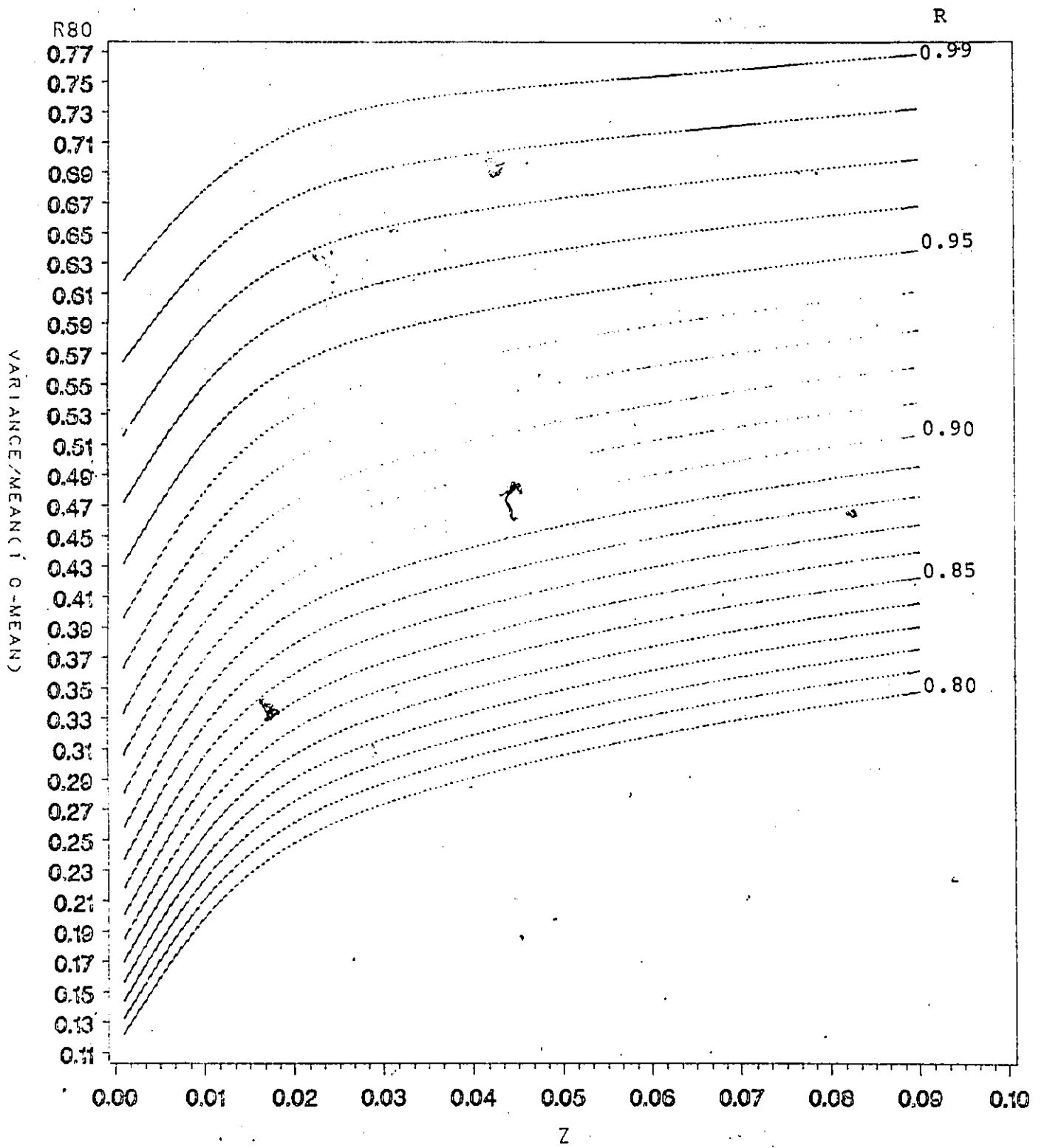


WITH SPLINE FITTING

351



WITH SPLINE FITTING

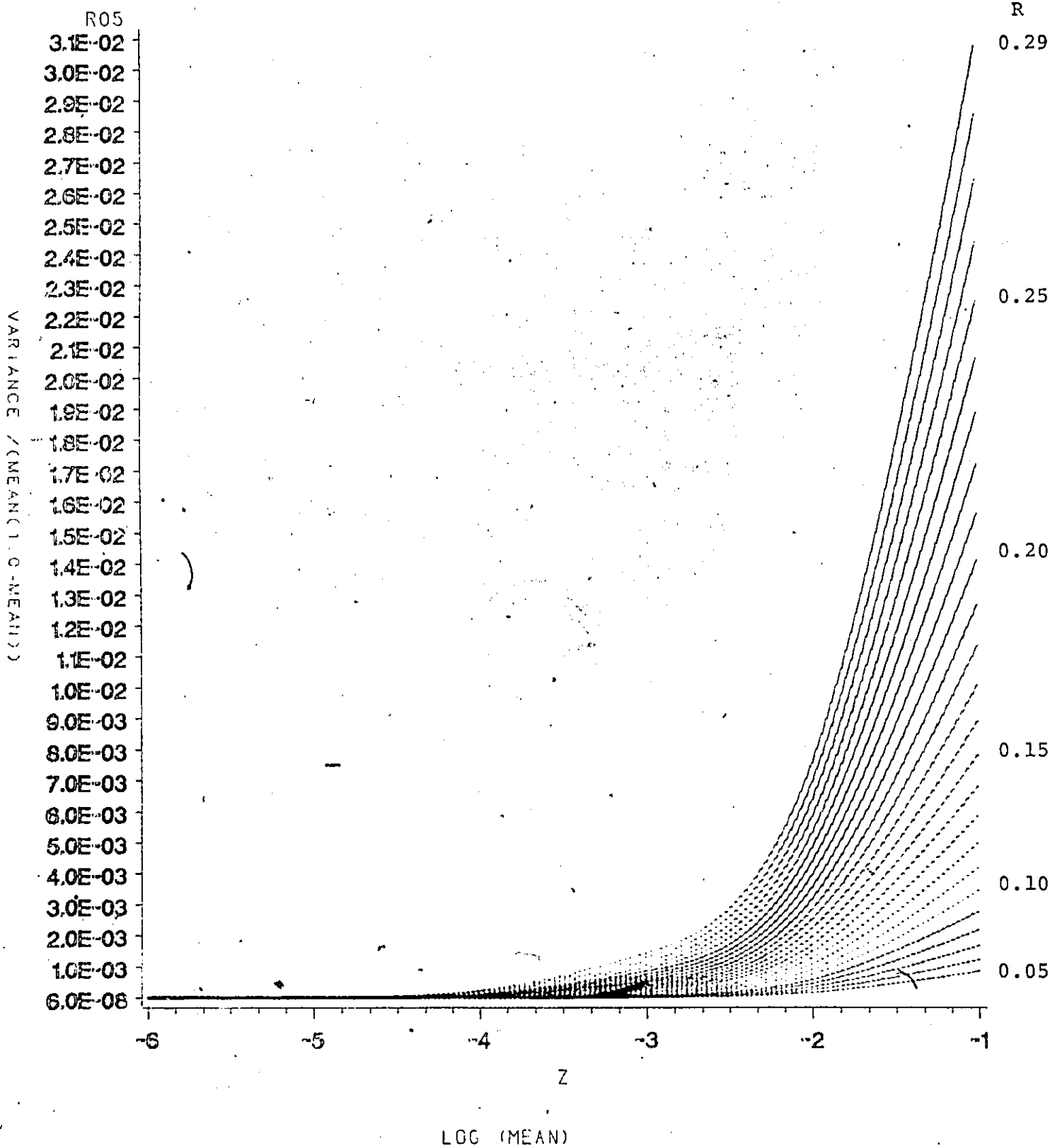


MEAN

3 - Relationships between logarithmic mean and variance/  
(mean(1.0 - mean))

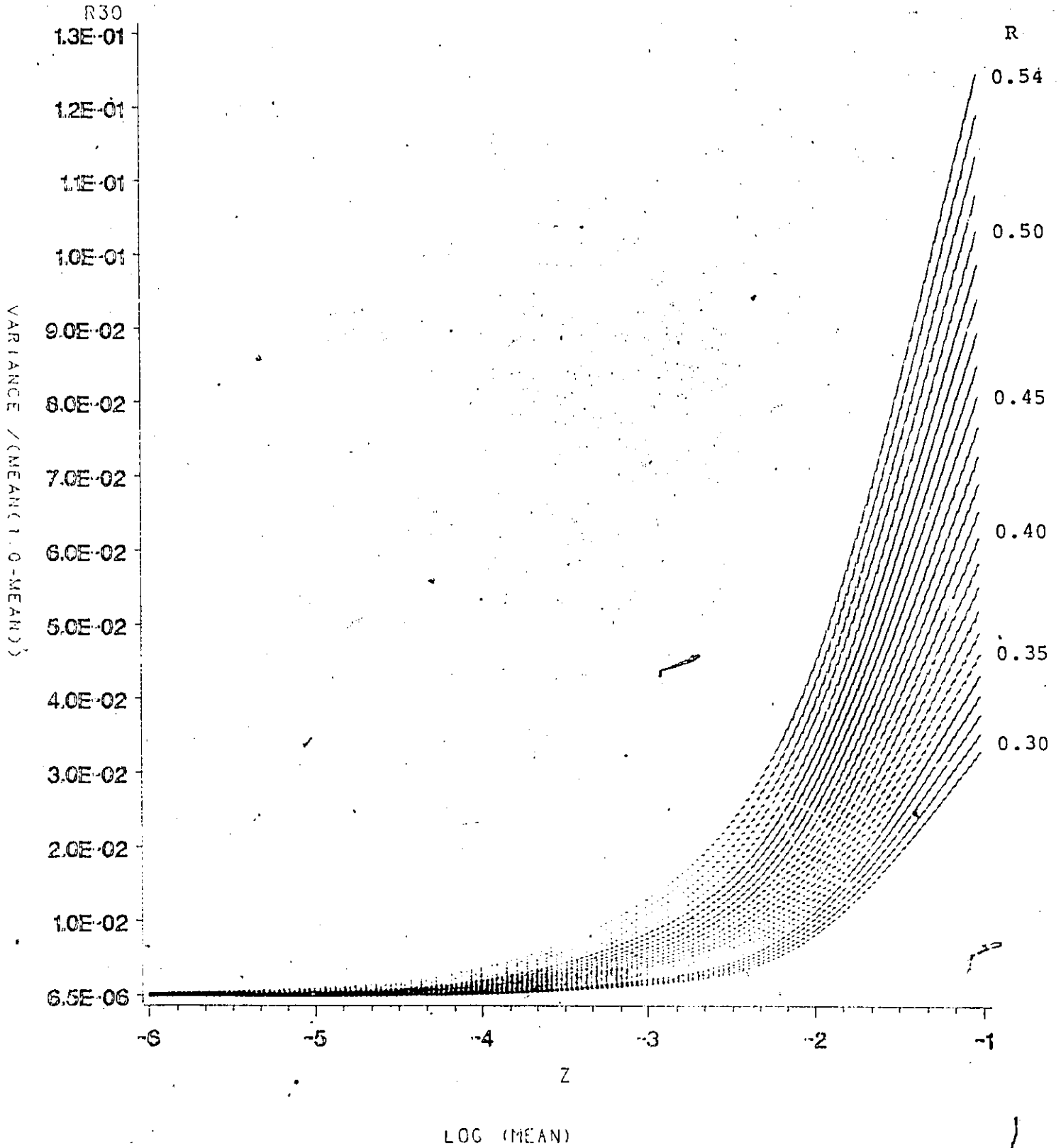
WITH SPLINE FITTING

354



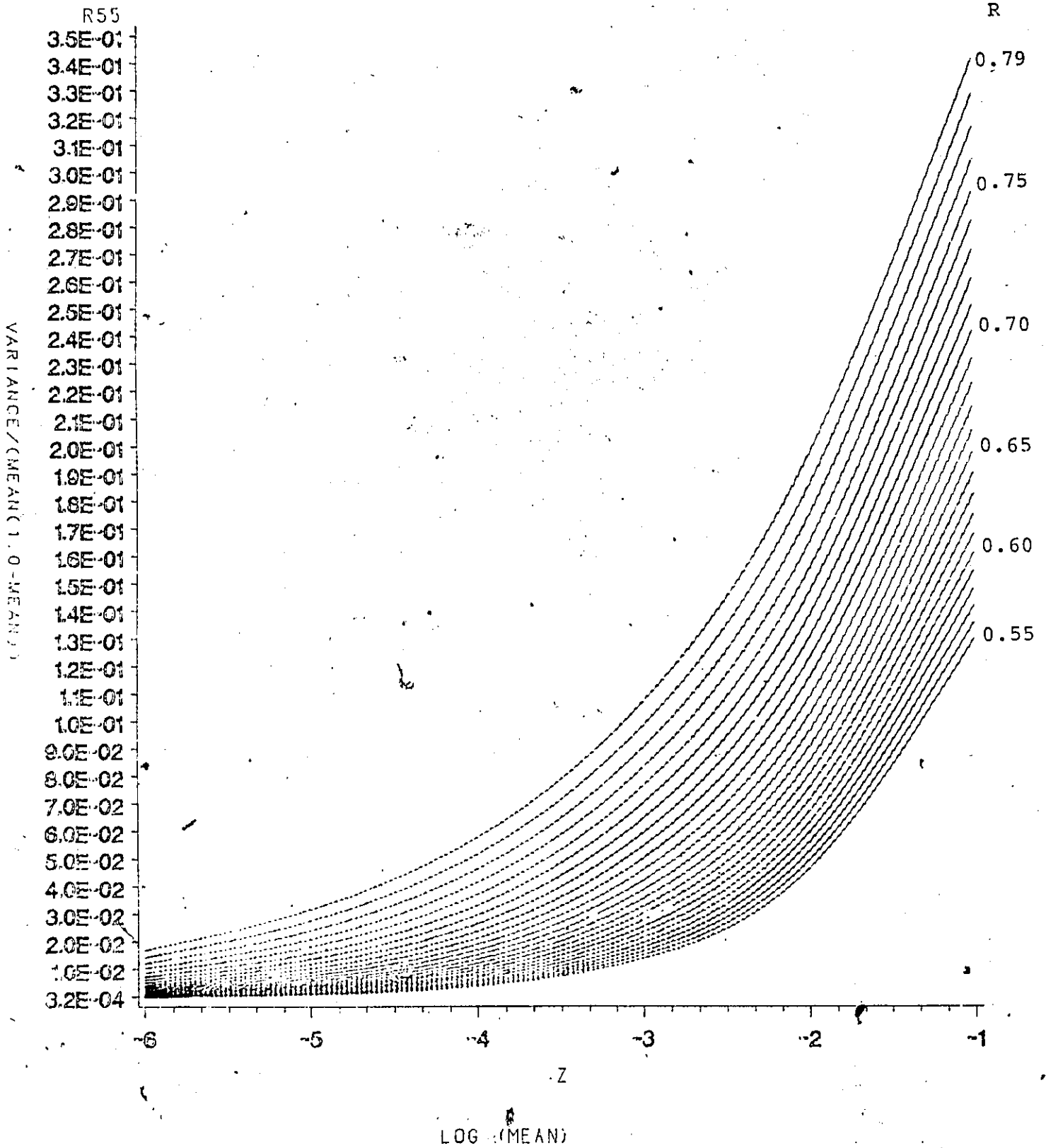
WITH SPLINE FITTING

355

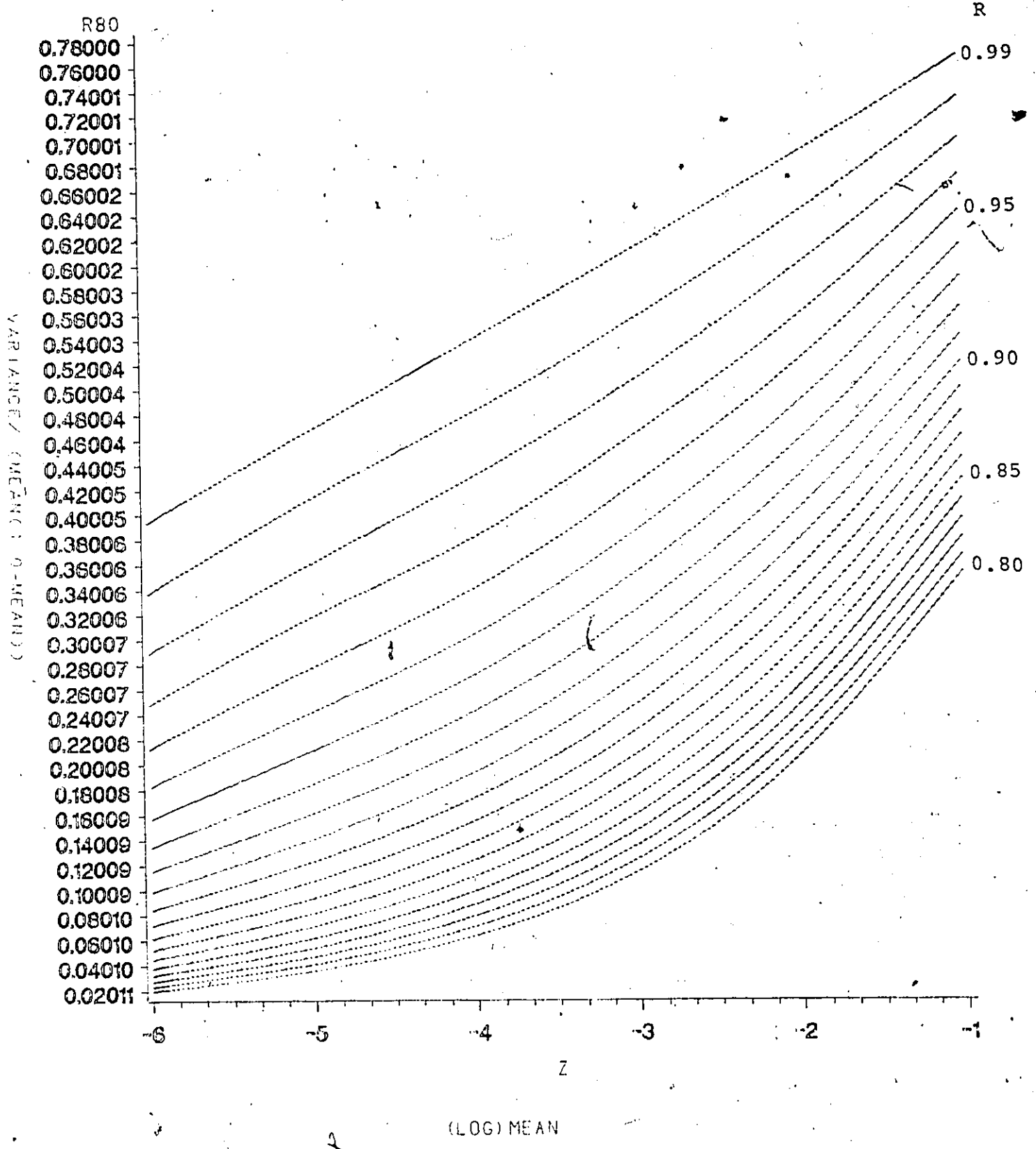


WITH SPLINE FITTING

356



WITH SPLINE FITTING



APPENDIX III: Relationships between mean and variance for  
various R values (TABLES)

- 1 - Relationships between mean and variance
- 2 - Relationships between mean and variance/(mean  
(1.0 - mean))
- 3 - Relationships between logarithmic mean and variance/  
(mean(1.0 - mean))

1 - Relationships between mean and variance

Relationships between mean and variance

R	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.05	0.000000	0.000002	0.000006	0.000012	0.000019	0.000027	0.000036	0.000045	0.000055	0.000066
0.06	0.000000	0.000003	0.000008	0.000017	0.000027	0.000038	0.000051	0.000065	0.000080	0.000095
0.07	0.000000	0.000004	0.000012	0.000023	0.000037	0.000052	0.000070	0.000089	0.000109	0.000130
0.08	0.000000	0.000005	0.000015	0.000030	0.000048	0.000069	0.000091	0.000116	0.000142	0.000170
0.09	0.000000	0.000006	0.000019	0.000038	0.000061	0.000087	0.000116	0.000147	0.000180	0.000215
0.10	0.000000	0.000007	0.000024	0.000047	0.000075	0.000108	0.000144	0.000182	0.000223	0.000266
0.11	0.000000	0.000009	0.000029	0.000057	0.000091	0.000131	0.000174	0.000221	0.000271	0.000322
0.12	0.000000	0.000011	0.000035	0.000068	0.000109	0.000156	0.000208	0.000264	0.000323	0.000384
0.13	0.000000	0.000013	0.000041	0.000085	0.000129	0.000184	0.000245	0.000310	0.000380	0.000452
0.14	0.000000	0.000015	0.000048	0.000094	0.000150	0.000214	0.000285	0.000361	0.000442	0.000526
0.15	0.000000	0.000017	0.000055	0.000108	0.000173	0.000247	0.000328	0.000361	0.000442	0.000526
0.16	0.000000	0.000020	0.000063	0.000124	0.000197	0.000282	0.000374	0.000474	0.000580	0.000690
0.17	0.000000	0.000022	0.000072	0.000141	0.000224	0.000319	0.000424	0.000537	0.000657	0.000781
0.18	0.000000	0.000025	0.000081	0.000159	0.000252	0.000360	0.000478	0.000604	0.000739	0.000879
0.19	0.000000	0.000028	0.000091	0.000178	0.000283	0.000403	0.000534	0.000676	0.000827	0.000982
0.20	0.000000	0.000032	0.000102	0.000198	0.000315	0.000449	0.000595	0.000752	0.000919	0.001092
0.21	0.000000	0.000035	0.000113	0.000220	0.000350	0.000497	0.000659	0.000833	0.001018	0.001209
0.22	0.000000	0.000039	0.000125	0.000244	0.000386	0.000549	0.000727	0.000918	0.001121	0.001331
0.23	0.000000	0.000043	0.000138	0.000268	0.000425	0.000603	0.000799	0.001009	0.001231	0.001461
0.24	0.000000	0.000048	0.000152	0.000295	0.000466	0.000661	0.000875	0.001104	0.001346	0.001597
0.25	0.000000	0.000053	0.000167	0.000322	0.000509	0.000722	0.000954	0.001204	0.001468	0.001740
0.26	0.000000	0.000058	0.000182	0.000352	0.000555	0.000786	0.001038	0.001309	0.001595	0.001891
0.27	0.000000	0.000063	0.000199	0.000383	0.000603	0.000854	0.001127	0.001419	0.001729	0.002048
0.28	0.000000	0.000069	0.000216	0.000415	0.000654	0.000924	0.001219	0.001535	0.001869	0.002213
0.29	0.000000	0.000075	0.000234	0.000450	0.000707	0.000999	0.001317	0.001657	0.002016	0.002386
0.30	0.000000	0.000081	0.000254	0.000486	0.000764	0.001077	0.001419	0.001784	0.002169	0.002566
0.31	0.000000	0.000088	0.000274	0.000525	0.000823	0.001159	0.001525	0.001916	0.002329	0.002754
0.32	0.000000	0.000096	0.000296	0.000565	0.000884	0.001245	0.001637	0.002055	0.002496	0.002950
0.33	0.000000	0.000103	0.000319	0.000607	0.000949	0.001335	0.001754	0.002200	0.002671	0.003155
0.34	0.000000	0.000112	0.000343	0.000652	0.001018	0.001429	0.001876	0.002352	0.002853	0.003368
0.35	0.000000	0.000120	0.000369	0.000699	0.001089	0.001528	0.002003	0.002510	0.003042	0.003590
0.36	0.000000	0.000130	0.000395	0.000748	0.001164	0.001631	0.002136	0.002674	0.003240	0.003820
0.37	0.000000	0.000140	0.000424	0.000800	0.001242	0.001739	0.002275	0.002846	0.003445	0.004060
0.38	0.000000	0.000150	0.000454	0.000854	0.001324	0.001851	0.002420	0.003024	0.003659	0.004309
0.39	0.000000	0.000161	0.000485	0.000911	0.001410	0.001968	0.002571	0.003210	0.003881	0.004568
0.40	0.000000	0.000173	0.000518	0.000970	0.001500	0.002091	0.002728	0.003403	0.004112	0.004837
0.41	0.000000	0.000186	0.000553	0.001033	0.001594	0.002216	0.002892	0.003604	0.004351	0.005116
0.42	0.000000	0.000199	0.000590	0.001098	0.001692	0.002352	0.003062	0.003814	0.004600	0.005405

MEAN

R	X	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.43	0.000000	0.000213	0.000628	0.001167	0.001794	0.002491	0.003240	0.004031	0.004859	0.005705	
0.44	0.000000	0.000228	0.000669	0.001239	0.001901	0.002636	0.003424	0.004257	0.005127	0.006016	
0.45	0.000000	0.000244	0.000712	0.001315	0.002013	0.002787	0.003616	0.004491	0.005405	0.006338	
0.46	0.000000	0.000261	0.000757	0.001393	0.002130	0.002944	0.003816	0.004735	0.005693	0.006672	
0.47	0.000000	0.000278	0.000804	0.001476	0.002252	0.003108	0.004023	0.004987	0.005993	0.007017	
0.48	0.000000	0.000297	0.000853	0.001563	0.002379	0.003279	0.004239	0.005250	0.006303	0.007375	
0.49	0.000000	0.000317	0.000906	0.001653	0.002512	0.003456	0.004464	0.005522	0.006624	0.007745	
0.50	0.000000	0.000338	0.000961	0.001748	0.002650	0.003641	0.004697	0.005805	0.006957	0.008129	
0.51	0.000000	0.000360	0.001018	0.001847	0.002795	0.003834	0.004939	0.006098	0.007302	0.008526	
0.52	0.000000	0.000384	0.001079	0.001951	0.002945	0.004034	0.005190	0.006402	0.007659	0.008936	
0.53	0.000000	0.000409	0.001142	0.002060	0.003102	0.004242	0.005451	0.006717	0.008029	0.009361	
0.54	0.000000	0.000435	0.001209	0.002173	0.003266	0.004459	0.005723	0.007043	0.008412	0.009800	
0.55	0.000000	0.000463	0.001280	0.002292	0.003437	0.004685	0.006004	0.007382	0.008809	0.010254	
0.56	0.000000	0.000493	0.001353	0.002416	0.003615	0.004919	0.006297	0.007733	0.009219	0.010723	
0.57	0.000000	0.000525	0.001431	0.002546	0.003801	0.005163	0.006600	0.008097	0.009644	0.011209	
0.58	0.000000	0.000558	0.001512	0.002682	0.003995	0.005417	0.006915	0.008474	0.010084	0.011711	
0.59	0.000000	0.000593	0.001598	0.002824	0.004197	0.005681	0.007242	0.008865	0.010539	0.012230	
0.60	0.000000	0.000630	0.001688	0.002972	0.004407	0.005956	0.007582	0.009271	0.011011	0.012767	
0.61	0.000000	0.000670	0.001782	0.003127	0.004626	0.006241	0.007934	0.009691	0.011499	0.013321	
0.62	0.000000	0.000711	0.001881	0.003290	0.004855	0.006538	0.008300	0.010126	0.012003	0.013895	
0.63	0.000000	0.000756	0.001985	0.003459	0.005093	0.006847	0.008680	0.010577	0.012526	0.014488	
0.64	0.000000	0.000802	0.002094	0.003637	0.005342	0.007168	0.009074	0.011045	0.013067	0.015101	
0.65	0.000000	0.000852	0.002209	0.003822	0.005601	0.007502	0.009484	0.011530	0.013627	0.015735	
0.66	0.000000	0.000904	0.002330	0.004017	0.005871	0.007849	0.009909	0.012032	0.014207	0.016391	
0.67	0.000000	0.000960	0.002456	0.004220	0.006153	0.008211	0.010350	0.012553	0.014808	0.017069	
0.68	0.000000	0.001018	0.002589	0.004432	0.006446	0.008587	0.010809	0.013094	0.015429	0.017771	
0.69	0.000000	0.001080	0.002729	0.004654	0.006753	0.008978	0.011285	0.013654	0.016074	0.018497	
0.70	0.000000	0.001146	0.002876	0.004887	0.007072	0.009386	0.011780	0.014235	0.016741	0.019249	
0.71	0.000000	0.001216	0.003030	0.005130	0.007406	0.009810	0.012294	0.014839	0.017433	0.020027	
0.72	0.000000	0.001290	0.003193	0.005385	0.007754	0.010252	0.012828	0.015465	0.018150	0.020832	
0.73	0.000000	0.001368	0.003363	0.005651	0.008118	0.010712	0.013384	0.016116	0.018894	0.021667	
0.74	0.000000	0.001451	0.003543	0.005931	0.008497	0.011192	0.013962	0.016791	0.019665	0.022531	
0.75	0.000000	0.001539	0.003732	0.006224	0.008894	0.011692	0.014564	0.017493	0.020466	0.023428	
0.76	0.000000	0.001632	0.003931	0.006531	0.009309	0.012213	0.015191	0.018223	0.021297	0.024358	
0.77	0.000000	0.001730	0.004141	0.006853	0.009743	0.012758	0.015843	0.018982	0.022161	0.025322	
0.78	0.000000	0.001835	0.004362	0.007191	0.010197	0.013326	0.016524	0.019772	0.023058	0.026324	
0.79	0.000000	0.001947	0.004594	0.007546	0.010672	0.013920	0.017233	0.020595	0.023992	0.027365	
0.80	0.000000	0.002065	0.004840	0.007919	0.011170	0.014541	0.017974	0.021452	0.024964	0.028448	

MEAN

$\bar{X}$	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.81	0.000000	0.002191	0.005099	0.008311	0.011694	0.015190	0.018747	0.022346	0.025976	0.029575
0.82	0.000000	0.002324	0.005374	0.008724	0.012240	0.015871	0.019556	0.023280	0.027032	0.030748
0.83	0.000000	0.002466	0.005663	0.009159	0.012816	0.016583	0.020401	0.024255	0.028134	0.031971
0.84	0.000000	0.002618	0.005970	0.009618	0.013422	0.017331	0.021287	0.025275	0.029284	0.033248
0.85	0.000000	0.002780	0.006295	0.010102	0.014059	0.018117	0.022216	0.026343	0.030488	0.034581
0.86	0.000000	0.002952	0.006640	0.010645	0.014731	0.018943	0.023191	0.027462	0.031747	0.035977
0.87	0.000000	0.003136	0.007006	0.011155	0.015439	0.019813	0.024216	0.028637	0.033068	0.037438
0.88	0.000000	0.003333	0.007396	0.011729	0.016189	0.020730	0.025294	0.029871	0.034455	0.038970
0.89	0.000000	0.003544	0.007810	0.012338	0.016981	0.021699	0.026431	0.031171	0.035913	0.040581
0.90	0.000000	0.003769	0.008253	0.012986	0.017822	0.022723	0.027631	0.032541	0.037448	0.042275
0.91	0.000000	0.004012	0.008726	0.013676	0.018715	0.023809	0.028901	0.033988	0.039067	0.044061
0.92	0.000000	0.004273	0.009232	0.014412	0.019666	0.024961	0.030245	0.035518	0.040779	0.045947
0.93	0.000000	0.004553	0.009775	0.015199	0.020679	0.026187	0.031673	0.037141	0.042592	0.047943
0.94	0.000000	0.004855	0.010358	0.016044	0.021762	0.027494	0.033192	0.038864	0.044515	0.050060
0.95	0.000000	0.005182	0.010987	0.016951	0.022923	0.028890	0.034811	0.040699	0.046560	0.052310
0.96	0.000000	0.005536	0.011665	0.017928	0.024169	0.030385	0.036541	0.042656	0.048741	0.054708
0.97	0.000000	0.005919	0.012400	0.018983	0.025511	0.031990	0.038394	0.044749	0.051070	0.057268
0.98	0.000000	0.006335	0.013197	0.020126	0.026960	0.033718	0.040384	0.046993	0.053565	0.060009
0.99	0.000000	0.006787	0.014064	0.021367	0.028529	0.035582	0.042526	0.049405	0.056245	0.062953

2

2 - Relationships between mean and variance/(mean(1.0 -  
mean ))

Relationships between mean and variance/(mean(t-0-mean))

R	0.001	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.05	0.000029	0.000181	0.000300	0.000399	0.000485	0.000561	0.000630	0.000694	0.000753	0.000806
0.06	0.000042	0.000261	0.000433	0.000576	0.000699	0.000810	0.000909	0.001000	0.001085	0.001162
0.07	0.000057	0.000357	0.000592	0.000786	0.000954	0.001104	0.001239	0.001364	0.001479	0.001584
0.08	0.000075	0.000468	0.000775	0.001029	0.001248	0.001445	0.001622	0.001784	0.001934	0.002071
0.09	0.000096	0.000595	0.000984	0.001306	0.001584	0.001833	0.002057	0.002262	0.002452	0.002625
0.10	0.000119	0.000738	0.001220	0.001618	0.001961	0.002269	0.002545	0.002798	0.003033	0.003247
0.11	0.000146	0.000898	0.001483	0.001965	0.002390	0.002753	0.003087	0.003393	0.003678	0.003936
0.12	0.000175	0.001076	0.001773	0.002348	0.002843	0.003286	0.003684	0.004048	0.004387	0.004694
0.13	0.000208	0.001271	0.002092	0.002768	0.003350	0.003870	0.004336	0.004764	0.005161	0.005521
0.14	0.000245	0.001485	0.002440	0.003225	0.003901	0.004504	0.005045	0.005541	0.006001	0.006418
0.15	0.000285	0.001718	0.002818	0.003721	0.004498	0.005191	0.005812	0.006380	0.006909	0.007387
0.16	0.000329	0.001971	0.003227	0.004257	0.005142	0.005930	0.006637	0.007284	0.007884	0.008428
0.17	0.000377	0.002245	0.003669	0.004833	0.005833	0.006724	0.007522	0.008252	0.008929	0.009542
0.18	0.000430	0.002541	0.004143	0.005452	0.006574	0.007574	0.008468	0.009286	0.010044	0.010730
0.19	0.000488	0.002859	0.004552	0.006173	0.007366	0.008445	0.009477	0.010387	0.011231	0.011995
0.20	0.000551	0.003201	0.005196	0.006820	0.008209	0.009445	0.010549	0.011557	0.012492	0.013336
0.21	0.000619	0.003568	0.005777	0.007520	0.009107	0.010469	0.011686	0.012797	0.013826	0.014756
0.22	0.000694	0.003961	0.006397	0.008372	0.010059	0.011555	0.012891	0.014109	0.015237	0.016256
0.23	0.000775	0.004381	0.007057	0.009222	0.011068	0.012704	0.014164	0.015494	0.016726	0.017837
0.24	0.000863	0.004830	0.007758	0.010123	0.012135	0.013918	0.015507	0.016954	0.018293	0.019501
0.25	0.000959	0.005309	0.008503	0.011076	0.013264	0.015199	0.016922	0.018491	0.019942	0.021250
0.26	0.001063	0.005821	0.009293	0.012085	0.014454	0.016548	0.018412	0.020106	0.021674	0.023085
0.27	0.001176	0.006365	0.010131	0.013150	0.015709	0.017969	0.019977	0.021803	0.023491	0.025009
0.28	0.001299	0.006945	0.011017	0.014275	0.017031	0.019462	0.021621	0.023582	0.025394	0.027023
0.29	0.001432	0.007562	0.011956	0.015461	0.018422	0.021031	0.023345	0.025447	0.027386	0.029130
0.30	0.001577	0.008219	0.012948	0.016711	0.019885	0.022677	0.025153	0.027398	0.029470	0.031331
0.31	0.001734	0.008917	0.013996	0.018027	0.021421	0.024404	0.027045	0.029439	0.031647	0.033628
0.32	0.001904	0.009658	0.015103	0.019412	0.023034	0.026213	0.029025	0.031572	0.033919	0.036025
0.33	0.002089	0.010446	0.016272	0.020869	0.024727	0.028108	0.031096	0.033800	0.036290	0.038522
0.34	0.002290	0.011283	0.017505	0.022401	0.026501	0.030091	0.033260	0.036125	0.038762	0.041124
0.35	0.002508	0.012171	0.018806	0.024011	0.028362	0.032165	0.035520	0.038550	0.041337	0.043831
0.36	0.002745	0.013113	0.020178	0.025702	0.030310	0.034334	0.037878	0.041078	0.044018	0.046648
0.37	0.003003	0.014113	0.021623	0.027477	0.032351	0.036600	0.040339	0.043711	0.046809	0.049576
0.38	0.003282	0.015174	0.023146	0.029340	0.034486	0.038968	0.042906	0.046454	0.049711	0.052619
0.39	0.003586	0.016300	0.024750	0.031294	0.036721	0.041439	0.045580	0.049309	0.052728	0.055779
0.40	0.003916	0.017493	0.026439	0.033344	0.039058	0.044019	0.048367	0.052279	0.055864	0.059059
0.41	0.004274	0.018758	0.028217	0.035493	0.041501	0.046710	0.051270	0.055368	0.059121	0.062464
0.42	0.004664	0.020099	0.030088	0.037746	0.044055	0.049517	0.054293	0.058580	0.062504	0.065995

R	$\bar{X}$	0.001	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.43	0.005087	0.021520	0.032057	0.040107	0.046724	0.052443	0.057438	0.061918	0.066015	0.069657	
0.44	0.005547	0.023026	0.034129	0.042581	0.049512	0.055494	0.060711	0.065386	0.069658	0.073452	
0.45	0.006048	0.024622	0.036308	0.045171	0.052424	0.058673	0.064116	0.068988	0.073438	0.077385	
0.46	0.006592	0.026313	0.038600	0.047884	0.055464	0.061984	0.067657	0.072729	0.077357	0.081460	
0.47	0.007183	0.028105	0.041010	0.050725	0.058638	0.065434	0.071338	0.076612	0.081421	0.085680	
0.48	0.007826	0.030002	0.043543	0.053699	0.061950	0.069026	0.075165	0.080643	0.085634	0.090049	
0.49	0.008525	0.032012	0.046207	0.056811	0.065407	0.072766	0.079142	0.084825	0.089999	0.094572	
0.50	0.009286	0.034141	0.049006	0.060069	0.069014	0.076659	0.083274	0.089164	0.094523	0.099253	
0.51	0.010112	0.036395	0.051947	0.063477	0.072776	0.080711	0.087567	0.093665	0.099209	0.104097	
0.52	0.011011	0.038782	0.055039	0.067042	0.076701	0.084928	0.092026	0.098333	0.104062	0.109108	
0.53	0.011989	0.041309	0.058286	0.070772	0.080793	0.089315	0.096657	0.103174	0.109088	0.114292	
0.54	0.013052	0.043984	0.061699	0.074674	0.085063	0.093880	0.101466	0.108193	0.114293	0.119654	
0.55	0.014208	0.046817	0.065283	0.078755	0.089512	0.098629	0.106459	0.113396	0.119681	0.125199	
0.56	0.015464	0.049815	0.069048	0.083022	0.094152	0.103568	0.111644	0.118790	0.125260	0.130933	
0.57	0.016831	0.052989	0.073003	0.087485	0.098989	0.108705	0.117026	0.124381	0.131034	0.136862	
0.58	0.018317	0.056348	0.077157	0.092152	0.104031	0.114048	0.122613	0.130176	0.137011	0.142992	
0.59	0.019933	0.059904	0.081520	0.097031	0.109288	0.119605	0.128413	0.136182	0.143198	0.149330	
0.60	0.021689	0.063667	0.086102	0.102133	0.114768	0.125384	0.134433	0.142406	0.149602	0.155882	
0.61	0.023599	0.067650	0.090914	0.107468	0.120479	0.131393	0.140682	0.148858	0.156230	0.162656	
0.62	0.025675	0.071865	0.095967	0.113046	0.126434	0.137644	0.147169	0.155544	0.163091	0.169660	
0.63	0.027931	0.076326	0.101275	0.118878	0.132640	0.144144	0.153903	0.162474	0.170192	0.176901	
0.64	0.030385	0.081048	0.106849	0.124977	0.139111	0.150904	0.160894	0.169658	0.177543	0.184388	
0.65	0.033051	0.086045	0.112704	0.131355	0.145657	0.157936	0.168152	0.177104	0.185153	0.192131	
0.66	0.035950	0.091335	0.118854	0.138025	0.152890	0.165251	0.175688	0.184824	0.193032	0.201138	
0.67	0.039100	0.096934	0.125314	0.145001	0.160224	0.172861	0.183514	0.192828	0.201190	0.208420	
0.68	0.042525	0.102861	0.132102	0.152298	0.167873	0.180779	0.191641	0.201129	0.209640	0.216987	
0.69	0.046247	0.109137	0.139234	0.159933	0.175852	0.189020	0.200084	0.209738	0.218393	0.225852	
0.70	0.050293	0.115782	0.146729	0.167922	0.184176	0.197597	0.208857	0.218670	0.227461	0.235027	
0.71	0.054691	0.122819	0.154607	0.176284	0.192862	0.206528	0.217974	0.227938	0.236859	0.244525	
0.72	0.059472	0.130273	0.162891	0.185039	0.201929	0.215829	0.227451	0.237559	0.246603	0.254361	
0.73	0.064670	0.138170	0.171602	0.194206	0.211396	0.225519	0.237307	0.247549	0.256706	0.264549	
0.74	0.070322	0.146539	0.180767	0.203811	0.221284	0.235617	0.247560	0.257925	0.267188	0.275108	
0.75	0.076467	0.155410	0.190411	0.213875	0.231617	0.246144	0.258229	0.268708	0.278066	0.286054	
0.76	0.083152	0.164817	0.200564	0.224427	0.242418	0.257125	0.269338	0.279917	0.289362	0.297407	
0.77	0.090423	0.174795	0.211259	0.235496	0.253716	0.268585	0.280911	0.291577	0.301095	0.309189	
0.78	0.098336	0.185385	0.222528	0.247111	0.265537	0.280550	0.292973	0.303713	0.313293	0.321424	
0.79	0.106948	0.196628	0.234409	0.259310	0.277916	0.293051	0.305553	0.316351	0.325980	0.334135	
0.80	0.116326	0.208571	0.246945	0.272128	0.290887	0.306121	0.318682	0.329522	0.339186	0.347353	

R	$\bar{X}$	0.001	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.81	0.126540	0.221265	0.260179	0.285608	0.304488	0.319796	0.332395	0.343259	0.352943	0.361109	
0.82	0.137671	0.234766	0.274162	0.299794	0.318761	0.334116	0.346729	0.357598	0.367285	0.375436	
0.83	0.149806	0.249135	0.288949	0.314739	0.333756	0.349125	0.361727	0.372579	0.382251	0.390371	
0.84	0.163045	0.264441	0.304601	0.330498	0.349522	0.364872	0.377434	0.388245	0.397885	0.405958	
0.85	0.177496	0.280757	0.321185	0.347133	0.366119	0.381411	0.393902	0.404648	0.414234	0.422242	
0.86	0.193281	0.298168	0.338776	0.364715	0.383611	0.398803	0.411188	0.421841	0.431351	0.439276	
0.87	0.210537	0.316764	0.357460	0.383322	0.402071	0.417116	0.429357	0.439886	0.449296	0.457118	
0.88	0.229417	0.336649	0.377329	0.403043	0.421580	0.436425	0.448480	0.458851	0.468135	0.475832	
0.89	0.250093	0.357935	0.398492	0.423976	0.442230	0.456816	0.468636	0.478813	0.487943	0.495492	
0.90	0.272758	0.380751	0.421066	0.446233	0.464125	0.478384	0.489915	0.499857	0.508802	0.516178	
0.91	0.297630	0.405240	0.445188	0.469943	0.487380	0.501238	0.512421	0.522081	0.530808	0.537986	
0.92	0.324956	0.431561	0.471011	0.495246	0.512130	0.525501	0.536266	0.545594	0.554064	0.561018	
0.93	0.355014	0.459895	0.498709	0.522311	0.538525	0.551308	0.561579	0.570517	0.578691	0.585390	
0.94	0.388123	0.490445	0.528479	0.551322	0.566735	0.578819	0.588506	0.596990	0.604825	0.611238	
0.95	0.424640	0.523438	0.560548	0.582493	0.596957	0.608211	0.617213	0.625171	0.632617	0.638712	
0.96	0.464978	0.559134	0.595172	0.616070	0.629414	0.639687	0.647888	0.655236	0.662240	0.667981	
0.97	0.509602	0.597824	0.632648	0.652334	0.664360	0.673478	0.680743	0.687389	0.693891	0.699244	
0.98	0.559048	0.639840	0.673312	0.691607	0.702089	0.709850	0.716021	0.721857	0.727791	0.732718	
0.99	0.613927	0.685557	0.717555	0.734261	0.742937	0.749102	0.753999	0.758903	0.764198	0.768657	

3 - Relationships between logarithmic mean and variance/  
(mean(1,0 - mean))

## Relationships between logarithmic mean and variance/(mean(1.0-mean))

R	$\bar{X}$	-1	-2	-3	-4	-5	-6
0.05	0.000000	0.000001	0.000004	0.000029	0.000181	0.000856	
0.06	0.000000	0.000001	0.000006	0.000042	0.000261	0.001234	
0.07	0.000000	0.000001	0.000008	0.000057	0.000357	0.001682	
0.08	0.000000	0.000001	0.000010	0.000075	0.000468	0.002199	
0.09	0.000000	0.000002	0.000013	0.000096	0.000595	0.002787	
0.10	0.000000	0.000002	0.000017	0.000119	0.000738	0.003447	
0.11	0.000000	0.000003	0.000021	0.000146	0.000898	0.004177	
0.12	0.000000	0.000003	0.000025	0.000175	0.001076	0.004981	
0.13	0.000000	0.000004	0.000030	0.000208	0.001271	0.005857	
0.14	0.000001	0.000005	0.000035	0.000245	0.001485	0.006808	
0.15	0.000001	0.000006	0.000041	0.000285	0.001718	0.007834	
0.16	0.000001	0.000007	0.000048	0.000329	0.001971	0.008936	
0.17	0.000001	0.000008	0.000055	0.000377	0.002245	0.010114	
0.18	0.000001	0.000009	0.000064	0.000430	0.002541	0.011371	
0.19	0.000001	0.000010	0.000073	0.000488	0.002859	0.012708	
0.20	0.000002	0.000012	0.000083	0.000551	0.003201	0.014125	
0.21	0.000002	0.000013	0.000094	0.000619	0.003568	0.015624	
0.22	0.000002	0.000015	0.000107	0.000694	0.003961	0.017206	
0.23	0.000002	0.000017	0.000121	0.000775	0.004381	0.018873	
0.24	0.000003	0.000020	0.000136	0.000863	0.004830	0.020627	
0.25	0.000003	0.000023	0.000153	0.000959	0.005309	0.022469	
0.26	0.000004	0.000026	0.000171	0.001063	0.005821	0.024400	
0.27	0.000004	0.000029	0.000192	0.001176	0.006365	0.026424	
0.28	0.000005	0.000033	0.000215	0.001299	0.006945	0.028540	
0.29	0.000006	0.000038	0.000240	0.001432	0.007562	0.030752	
0.30	0.000001	0.00004	0.00027	0.00158	0.00822	0.03306	
0.31	0.000001	0.00005	0.00030	0.00173	0.00892	0.03547	
0.32	0.000001	0.00006	0.00033	0.00190	0.00966	0.03798	
0.33	0.000001	0.00006	0.00037	0.00209	0.01045	0.04060	
0.34	0.000001	0.00007	0.00042	0.00229	0.01128	0.04332	
0.35	0.000001	0.00008	0.00046	0.00251	0.01217	0.04615	
0.36	0.000002	0.00009	0.00052	0.00275	0.01311	0.04909	
0.37	0.000002	0.00010	0.00057	0.00300	0.01411	0.05214	
0.38	0.000002	0.00012	0.00064	0.00328	0.01517	0.05531	
0.39	0.000002	0.00014	0.00071	0.00359	0.01630	0.05860	
0.40	0.000003	0.00015	0.00079	0.00392	0.01749	0.06202	
0.41	0.000003	0.00018	0.00088	0.00427	0.01876	0.06555	
0.42	0.000004	0.00020	0.00098	0.00466	0.02010	0.06922	
0.43	0.000005	0.00023	0.00110	0.00509	0.02152	0.07302	
0.44	0.000005	0.00026	0.00122	0.00555	0.02303	0.07695	
0.45	0.000006	0.00030	0.00136	0.00605	0.02462	0.08103	
0.46	0.000007	0.00034	0.00152	0.00659	0.02631	0.08524	
0.47	0.000009	0.00039	0.00169	0.00718	0.02810	0.08960	
0.48	0.000010	0.00044	0.00189	0.00783	0.03000	0.09412	
0.49	0.000012	0.00051	0.00211	0.00853	0.03201	0.09878	
0.50	0.000014	0.00058	0.00235	0.00929	0.03414	0.10361	
0.51	0.000016	0.00067	0.00262	0.01011	0.03659	0.10859	
0.52	0.000019	0.00076	0.00293	0.01101	0.03878	0.11375	
0.53	0.000023	0.00087	0.00327	0.01199	0.04131	0.11907	
0.54	0.000027	0.00100	0.00365	0.01305	0.04398	0.12458	
0.55	0.000032	0.00115	0.00407	0.01421	0.04682	0.13026	
0.56	0.000038	0.00132	0.00455	0.01546	0.04982	0.13613	
0.57	0.000044	0.00152	0.00508	0.01683	0.05299	0.14220	
0.58	0.000052	0.00174	0.00567	0.01832	0.05637	0.14847	

R	MEAN					
	$\bar{X}$	-1	-2	-3	-4	-5
0.59	0.00062	0.00200	0.00634	0.01993	0.05990	0.15494
0.60	0.00073	0.00230	0.00708	0.02169	0.06367	0.16163
0.61	0.00086	0.00264	0.00791	0.02360	0.06765	0.16853
0.62	0.00102	0.00303	0.00884	0.02567	0.07187	0.17566
0.63	0.00121	0.00348	0.00987	0.02793	0.07633	0.18303
0.64	0.00143	0.00400	0.01103	0.03038	0.08105	0.19063
0.65	0.00169	0.00460	0.01232	0.03305	0.08605	0.19849
0.66	0.00199	0.00528	0.01377	0.03595	0.09133	0.20661
0.67	0.00236	0.00607	0.01539	0.03910	0.09693	0.21500
0.68	0.00278	0.00697	0.01719	0.04252	0.10286	0.22368
0.69	0.00329	0.00801	0.01920	0.04625	0.10914	0.23264
0.70	0.00389	0.00920	0.02145	0.05029	0.11578	0.24191
0.71	0.00459	0.01057	0.02397	0.05469	0.12282	0.25149
0.72	0.00542	0.01214	0.02677	0.05947	0.13027	0.26141
0.73	0.00639	0.01394	0.02990	0.06467	0.13817	0.27167
0.74	0.00754	0.01600	0.03340	0.07032	0.14654	0.28230
0.75	0.00889	0.01836	0.03730	0.07647	0.15541	0.29330
0.76	0.01047	0.02107	0.04166	0.08315	0.16482	0.30470
0.77	0.01234	0.02418	0.04652	0.09042	0.17480	0.31653
0.78	0.01453	0.02773	0.05195	0.09834	0.18538	0.32879
0.79	0.01709	0.03181	0.05800	0.10695	0.19663	0.34152
0.80	0.02011	0.03647	0.06476	0.11633	0.20857	0.35475
0.81	0.02363	0.04181	0.07231	0.12654	0.22126	0.36850
0.82	0.02777	0.04792	0.08074	0.13767	0.23477	0.38282
0.83	0.03260	0.05491	0.09015	0.14981	0.24914	0.39772
0.84	0.03826	0.06291	0.10066	0.16304	0.26444	0.41327
0.85	0.04487	0.07207	0.11240	0.17750	0.28076	0.42950
0.86	0.05259	0.08254	0.12552	0.19328	0.29817	0.44646
0.87	0.06160	0.09452	0.14018	0.21054	0.31676	0.46421
0.88	0.07211	0.10823	0.15657	0.22942	0.33665	0.48282
0.89	0.08435	0.12391	0.17490	0.25009	0.35794	0.50236
0.90	0.09861	0.14185	0.19540	0.27276	0.38075	0.52290
0.91	0.11519	0.16237	0.21836	0.29763	0.40524	0.54455
0.92	0.13448	0.18585	0.24406	0.32496	0.43156	0.56739
0.93	0.15689	0.21270	0.27285	0.35501	0.45990	0.59156
0.94	0.18291	0.24344	0.30513	0.38812	0.49044	0.61719
0.95	0.21309	0.27862	0.34132	0.42464	0.52344	0.64442
0.96	0.24809	0.31889	0.38195	0.46498	0.55913	0.67342
0.97	0.28862	0.36499	0.42756	0.50960	0.59782	0.70440
0.98	0.33555	0.41778	0.47881	0.55905	0.63984	0.73757
0.99	0.38982	0.47826	0.53643	0.61393	0.68556	0.77319