

STOCHASTIC PROJECTIONS FOR WATER RESOURCES PLANNING

A Report by

Leo R. Beard, Technical Director

and

Shin Chang, Research Associate

Center for Research in Water Resources

The University of Texas at Austin

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Foreword

This is the final technical report issued under the Office of Water Research and Technology Grant No. 14-31-0001-9088, "Techniques for Projecting Alternative Futures for Water Resources Planning and for Estimating Flood Flow Frequencies". A previous report, "Flood Flow Frequency Techniques", was published on October 1, 1974.

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STOCHASTIC PROJECTIONS FOR WATER RESOURCES PLANNING

1. PROJECT OBJECTIVES

Evaluation of the future benefits and costs of water resources projections have traditionally been based on unique projections of key variables such as population and interest rates. Recognizing that the actual changes or trends that will occur in the future will be somewhat different from the theoretical projections, there is a fundamental need to evaluate the differences in benefits and costs that will occur if different trends occur. Also, the more comprehensive nature of water resources planning that is required today involves projections of many variables and, in projecting these variables, the interrelationships must be taken into consideration.

The objective of the research project reported on herein is to develop a technique for evaluating a range of projections of several types of variables that are involved in water resources planning and to develop techniques for using these projections effectively and efficiently. Variables used for illustration are economic, demographic and hydrologic. Only successive annual values are considered, since the project study is not concerned with seasonal variations. Only mathematical projection techniques are considered, since it is the technique of projecting sequences that is of concern to this project, and not developing cause-effect relationships wherein the projection of one variable can be used in deducing the projection of a second variable. Further, it is not an objective of this study to determine or develop best mathematical functions for projecting specific types of variables, but to suggest procedures, techniques and criteria for assessing the entire range of future potential as contrasted with the single most likely future.

2. BASIC CONSIDERATIONS

Of primary importance in most water resources planning studies, in

regard to projections of pertinent variables, is the projection level that will be reached at some specified time in the future. Of secondary importance are variations that occur along the line of projection. In most past studies, single projections of each pertinent variable have been made, and this ordinarily represents a smooth function that arrives at the most likely projection levels at various future times. Considering the uncertainty that available data on past quantities adequately represent future potential, and considering the virtual certainty that the actual future variations will not coincide with the most likely projections, water resources studies should consider alternative projection studies that can occur consistently with past variations of the particular variable being projected and considering our knowledge of how frequently and severely unexpected variations from indicated trends have occurred in the past.

As in the formulation of any mathematical model, the degree of complexity of a projection model must be adequate to represent variations of the phenomenon being studied with a degree of accuracy necessary for the particular applications anticipated. On the other hand, the model must be simple enough so that it can be calibrated satisfactorily with available data. Often it is not possible to meet both of these requirements, in which case the latter requirement would normally control, since there is little point in having a model that cannot be adequately calibrated. When this occurs, extreme caution must be used in interpreting the predictions that are made with the model.

Projection models described below have been restricted to a modest degree of complexity because of the limitations of data for calibration. Nevertheless, they include two types of variation that have not been used in traditional projection studies. These consist of a range of projections that adequately represent future possibilities and their associated probabilities, and variations within each projection that represent differences in trends that can be expected to occur from time to time. A small number of such projections covering the entire range of future expectations can be used separately to evaluate the accomplishments of proposed water resource projects, and expected project benefits can be computed by adding the cross products of the benefits obtained for each projection and the associated probability attached to that projection.

3. SINGLE-VARIABLE PROJECTION TECHNIQUES

Time series definitions. Some terms of special relevance to this report are defined as follows:

1. Time Series. A time series is a set of observations generated sequentially in time.
2. Stochastic Process. A stochastic process is a statistical phenomenon that evolves in time (or space) according to probabilistic laws. The time series to be analyzed may then be thought of as one particular realization, produced by the underlying probability mechanism, of the system under study.
3. Stationary Time Series. A time series is said to be stationary if the imbedded stochastic process of the time series remains in equilibrium about a constant level. Statistically, an nth-order stationary series requires that the first n moments of its distribution function be constant. For example, a second-order stationary series must have constant mean and constant variance.
4. Backward Shift Operator B. The backward shift operator B on a variable Z is defined by the equation: $B^m Z_t = Z_{t-m}$.
5. Backward Difference Operator V. The backward difference operator V on a variable Z is defined in terms of B as: $V = (1-B)$; hence $VZ_t = Z_t - Z_{t-1}$.
6. Differences of a Time Series. Differences of a time series are obtained by successively subtracting the previous value from the current one, and this can be repeated to obtain differences of higher order. The degree of differencing is designated d. Using the backward difference operator V, the dth difference of a time series Z can be represented as $V^d Z$. For example, $V^2 Z = V(VZ) = Z_t - 2Z_{t-1} + Z_{t-2}$.

Stationary Stochastic Models. Three categories of stationary stochastic models used in the description of stationary time series are autoregressive, moving average, and mixed autoregressive and moving average. These are defined as follows:

1. Autoregressive Models (AR). In these models, the current value

of the series is expressed as a finite, linear aggregate of the previous values plus a random component, a_t . Denoting the values of a time series at equally spaced times $t, t+1, \dots$, by Z_t, Z_{t+1}, \dots , and denoting the deviation from μ , the mean value of the time series, by $\bar{Z}_t, \dots, \bar{Z}_{t+1}$. Then the p -th AR model has the form

$$\begin{aligned} \mu &= \frac{1}{n} \sum_{t=1}^n Z_t \\ \bar{Z}_t &= \phi_1 \bar{Z}_{t-1} + \phi_2 \bar{Z}_{t-2} + \dots + \phi_p \bar{Z}_{t-p} + a_t \end{aligned} \quad (3.1)$$

where $\phi_1, \phi_2, \dots, \phi_p$, are p 's autoregressive parameters. The model contains $p+2$ unknown parameters, $\mu, \phi_1, \dots, \phi_p$ and σ_a^2 , to be identified from the data. The additional parameter σ_a^2 is the variance of the white noise process or random component, a_t . The stochastic AR model can be useful in the representation of certain practical series. For instance, the first order and second order Markov models frequently used in streamflow synthesis are AR models.

2. Moving Average Models (MA). In these models, the current value of the series, \bar{Z}_t , is expressed as a finite weighted sum of q previous random terms a_t . Thus the q -th MA model has the form

$$\bar{Z}_t = a_t - \theta_1 a_{t-1} - \theta_2 a_{t-2} - \dots - \theta_q a_{t-q} \quad (3.2)$$

where $\theta_1, \theta_2, \dots, \theta_q$ are q 's moving average parameters. This model contains $q + 2$ unknown parameters $\mu, \theta_1, \theta_2, \dots, \theta_q$ and σ_a^2 to be estimated from the data. Although this is called a moving average model, it is conceptually an infinite-term autoregressive model of specific form. The MA model is of practical importance in the representation of observed time series.

3. Mixed Autoregressive and Moving Average Models (ARMA). To achieve greater flexibility in fitting of actual time series, it is sometimes advantageous to include both autoregressive and moving average terms in one model. This leads to the mixed ARMA model. An ARMA(p, q) with p autoregressive terms and q moving average terms has the form,

$$\bar{Z}_t = \phi_1 \bar{Z}_{t-1} + \dots + \phi_p \bar{Z}_{t-p} + a_t - \phi_1 a_{t-1} - \dots - \phi_q a_{t-q} \quad (3.3)$$

The ARMA(p,q) model requires p+q+2 unknown parameters to be estimated from the data. It should be noted that the ARMA(p,q) model becomes an AR(p) model if q is set to zero and an MA(q) model if p is set to zero.

Nonstationary Stochastic Models. There are an unlimited number of ways in which a time series can be nonstationary. In fact, many time series in industry, business, and economics are better represented as nonstationary. Frequently, however, the nonstationary time series being analyzed exhibit a particular kind of homogenous nonstationary behavior. (i.e. the series exhibits homogeneity in the sense that, apart from the local level, or local level and trend, one segment of the series behaves much like any other segment). Differences of a nonstationary series of this kind can usually be expected to be stationary. Given the condition that the d-th difference of a nonstationary series is stationary, the autoregressive-integrated moving average model (which is a modified form of the ARMA model), can be used to describe the nonstationary series.

Autoregressive-Integrated Moving Average Models (ARIMA). The stochastic, nonstationary ARIMA model (with p autoregressive terms and q moving average terms) which describes a nonstationary series that is stationary in its dth difference has the form,

$$(1-\phi_1 B - \dots - \phi_p B^p) W_t = \theta_0 + (1-\theta_1 B - \dots - \theta_q B^q) a_t \quad (3.4)$$

with

$$W_t = \nabla^d Z_t, \text{ if } d > 0$$

$$= Z_t - \mu, \text{ if } d = 0$$

where Z is the series to be modeled,
 W is the d th difference of the series Z ,
 θ_0 is the long term trend factor.

It should be noted that the ARIMA(p, d, q) model becomes an AR(p) model if q and d are zero and an MA(q) model if p and d are zero and an ARMA model if d is zero.

Specific Models Studied. Seven models were selected for consideration under this project. They are

1. Second-Order Linear Markov Model.

$$Z_t = aC_1 Z_{t-1} + (1-a) C_2 Z_{t-2} \quad (3.5)$$

where a is the weighting factor, C_1 and C_2 are random variables;
 $C_1 = Z_t/Z_{t-1} + b_1 t$ and $C_2 = Z_t/Z_{t-2} + b_2 t$ and b_1 and b_2 are stochastic linear regression coefficients, and t is time. The values of b_1 and b_2 for forecasting are estimated stochastically as a function of Student's t distribution from the known historical data.

2. Second-Order Nonlinear Markov Model.

$$Z_t = Z_{t-1}^{aC_1} \cdot Z_{t-2}^{(1-a)C_2} \quad (3.6)$$

This model is the non-linear version of model 1. By taking logarithms of the variable, we have

$$\ln Z_t = aC_1 \ln Z_{t-1} + (1-a) C_2 \ln Z_{t-2}, \quad (3.7)$$

this equation has the same form as that of model 1 except that the logarithms of the time series Z are used instead of the time series itself. As in model 1, C_1 and C_2 are estimated stochastically.

3. Nth-Order Linear Markov Model.

$$Z_t = aC_1 Z_{t-1} + bC_2 Z_{t-2} + cC_3 Z_{t-3} + dC_4 Z_{t-4} + eC_5 Z_{t-5} \text{ for } N \leq 5. \quad (3.8)$$

The factors a, b, c, d, and e provide for weighting of the independent variables. The linear coefficients C_1 , C_2 , C_3 , C_4 , and C_5 are random variables and are distributed according to certain probability functions. For the purposes of this study, these coefficients were assumed to be normally distributed and defined by the equation

$$C_i = Z_t / Z_{t-i}, \quad i = 1, 2, 3, 4, 5 \quad (3.9)$$

When the model is being used for forecasting, the values of the linear coefficients are estimated stochastically from the historical data.

4. Nth-Order Nonlinear Markov Model.

$$Z_t = Z_{t-1}^{aC_1} \cdot Z_{t-2}^{bC_2} \cdot Z_{t-3}^{cC_3} \cdot Z_{t-4}^{dC_4} \cdot Z_{t-5}^{eC_5} \quad (3.10)$$

This model is the nonlinear version of model 3. By taking logarithms of the variable, the resulting equation has the same form as that of model 3.

5. Decade and Annual Ratio Autocorrelation Model. This model takes advantage of the availability of 10-year census data and the relatively accurate estimates of annual data that have been obtained using census techniques and omitting annual data obtained by mathematical interpolation. This permits the most effective use of early census information when only decade values are available. The model consists of a lag-1 Markov chain of the logarithms of the ratios of successive decade values and an interpolation routine using a similar Markov chain for annual logarithms where annual observations are available. When generating successive sequences, random values of average decade ratios are used for each sequence rather than sample values.

6. Decade Regression and Annual Autocorrelation Model. This model is similar to the preceding model except that a linear regression on the logarithms of the decade ratios is used instead of autocorrelation, and a new random value of the regression coefficient is used for each generated sequence.

7. The ARIMA(p,d,q). The ARIMA model includes three classes of models previously described: the AR, the MA, and the ARMA. It can be applied to both stationary and nonstationary time series. The degree of differencing, d , and the number of parameters, p and q , are identified first, then values of parameters are estimated by the least mean square errors method.

The first six models are essentially special cases of the AR process and they all experienced forecasting problems in some applications, particularly in the nature of forecasting unreasonable quantities. Thus, the ARIMA model was chosen as the standard forecasting model because of its flexibility and adaptability to various types of time series, thus providing means of avoiding unreasonable forecasts.

4. THE ARIMA MODEL

ARIMA(p,d, q) Model Building. To fit a series by an ARIMA(p,d,q) model, the model building process is ordinarily based on available observations. Computationally, the model building includes three steps: identification, estimation, and diagnostic checking. The order of differencing, d , the number of autoregressive parameters, p , and the number of moving average parameters, q , are identified in the identification step; values of parameters are then estimated in the estimation step; and the adequacy of the model in describing and projecting the series is checked in the diagnostic step.

According to Box and Jenkins (1970), no pure mathematical approach can be adopted in the model building process. Graphical methods and judgment are needed throughout because of the trial-and-error nature of the process. Following their basic ideas, a systematic noniterative algorithm is developed for model building, and although the algorithm is conceptually an exhaustive trial-and-error approach, operationally it is automatic. All possible models from the general ARIMA family of limited

order are examined, and the best model is selected such that four proposed performance criteria are either satisfied or maximized. These performance criteria are established to measure the goodness of fit and reasonableness of projections.

1. Identification of the Order of Differencing d. The failure of the autocorrelation function to die out quickly is taken as an indication that the underlying stochastic process of the series Z is nonstationary, but the process of its first or higher order differences may be stationary. Therefore, it is assumed that the degree of differencing, d, necessary to achieve stationarity has been reached when the autocorrelation function of $W_t = \nabla^d Z_t$ dies out quickly. In practice, d is generally 0, 1, or 2, and it is usually sufficient to inspect the first 20 (or so) estimated autocorrelations of the original series and of its first and second differences. According to Box and Jenkins (1970), to judge whether or not the autocorrelation function dies out quickly is completely subjective, as no standard judging criterion is available. The following method was found to be an effective way of detecting the speed of decay of the autocorrelation function and identifying the order of differencing, d.

Given a series of n elements and denoting the lag i autocorrelation coefficient by r_i , the Chi-square statistic, χ_d^2 , of the first 24 autocorrelation coefficients can be calculated as

$$\chi_d^2 = n \cdot \sum_{i=1}^{24} r_i^2, \quad (4.1)$$

Where d can be 0, 1, or 2 to stand for, respectively, the Chi-square statistic of the original series, of its first difference, or of its second difference. Let R_1 and R_2 be the ratios of

$$R_1 = \chi_1^2 / \chi_0^2, \text{ and } R_2 = \chi_2^2 / \chi_1^2 \quad (4.2)$$

In general, R_1 and R_2 are less than one. These ratios give the relative decay speed of the autocorrelation function of the series and its differences. For example, if $R_1 = 0.2$, then the Chi-square statistic of the

first difference is only 0.2 times that of the original series. This indicates that the speed of decay of the autocorrelation function of the first difference is five times faster than that of the original series. Based on this argument, the following rule can be used to select the degree of differencing d for the model:

$$\begin{aligned} \text{Select } d &= 0, \text{ if } R_1 > \epsilon \text{ and } R_2 > \epsilon; \\ d &= 1, \text{ if } R_1 < \epsilon \\ d &= 2, \text{ if } R_1 > \epsilon \text{ and } R_2 < \epsilon. \end{aligned}$$

The criterion, ϵ , is set from experience and judgment. A value of 0.4 was used in identifying d for all sample series in this study.

Parameter Estimation. Once it has been determined what the degree of differencing should be, values of parameters in all the member models of the general ARIMA family are next estimated. In the general order of decreasing complexity, the fourteen most used members of the ARIMA(p,d,q) family (listed by their (p,d,q) values) are: (2,d,3), (3,d,2), (1,d,3), (2,d,2), (3,d,1), (0,d,3), (1,d,2), (2,d,1), (3,d,0), (0,d,2), (1,d,1), (2,d,0), (0,d,1), (1,d,0). Box and Jenkins suggest that several of the models most likely to best describe the time series can be identified by comparing the shapes of their autocorrelation and partial autocorrelation functions with those of certain established theoretical patterns. The best model is then selected by a detailed study of the residuals. For the following reasons, however, it is believed that the following approach is more attractive and easier to apply than that of Box and Jenkins:

1. From the experience of modeling 30 sample series, it was found that in most cases, the autocorrelation and partial autocorrelation functions of the appropriately differenced series fail to show any pattern to yield a meaningful conclusion in selecting p and q in accordance with the findings of Box and Jenkins.

2. In some cases, models with different values of p and q can be fitted equally well to a given series and yield equally high levels of significance in the diagnostic checking step. This in turn suggests that

correspondence between the selection of p and q of an ARIMA(p,d,q) model and the shape of autocorrelation and partial autocorrelation functions is not unique.

3. There is no definite trend in either direction between the complexity (number of parameters used in a model) of a model and the performance (to be discussed in the model screening step) of the model. In other words, no selective search technique, rather than an exhaustive search of all possible models, can be developed to find the best model.

4. No plot of the autocorrelation and the partial autocorrelation function is required for the selected approach.

5. No intermediate subjective feedback is needed in the course of selecting the best model by the current approach; hence an automatic computer search procedure is possible.

The estimation of autoregressive and moving average parameters for an ARIMA(p,d,q) model is based on the Marquardt algorithm for nonlinear least-square parameter estimation. Computer programs based on this algorithm are available from the IMSL (International Mathematical and Statistical Libraries) program library 3 (for CDC Cyber 70/6000/7000), in which FTARPS and FTMAPS (both in FORTRAN) were used in this study to estimate the autoregressive and moving average parameters, respectively. After estimating the parameters of an ARIMA(p,d,q) model, the residuals, a_t , are computed by the equation

$$a_t = \hat{W}_t - W_t, \quad (4.3)$$

where \hat{W}_t is the fitted value of the time series and W_t the observed value.

Model Screening. Four criteria for selecting the best model to describe a time series are used:

1. Goodness of Fit. The goodness of fit of an ARIMA model to a time series is measured by the level of significance of the Chi-square test for the residual, a_t , in relation to white noise (randomness). The Chi-square statistic of the residual a_t for $t = 1, 2, \dots, n-d$, is calculated using the formula,

$$\chi^2 = (n-d) \sum_{i=1}^m r_i^2 \quad (4.4)$$

where r_i is the i th autocorrelation coefficient of series a_t , and m is the number of autocorrelation coefficients used in the computation of the statistic. With $m-p-q$ degrees of freedom, the significance level associated with χ^2 can be found from a Chi-square test table. The computer program MDCH of IMSL was used to find this level. The resulting residuals, a_t , should not significantly differ from a white noise process, if the time series of interest is fitted by an adequate ARIMA model. Thus, a minimum level of significance can be set to reject inadequate models. A minimum level of 0.01 was used in this study.

2. Model Stability. There exists an admissible region of parameters for every ARIMA(p,d,q) model which will ensure the stationarity and invertibility of the model (see chapter 3 of Box and Jenkins, 1970). Models with parameters outside this region will be dropped from further consideration. The region can be expressed implicitly as:

a. For Autoregressive Parameters. The roots of the polynomial $(1-\phi_1B-\phi_2B^2-\dots-\phi_pB^p)$ which is formed by taking all autoregressive of the ARIMA(p,d,q) model (i.e. left hand side of equation 3.4) when solved for B, must be greater than one.

b. For Moving Average Parameters. The roots of the polynomial $(1-\theta_1B-\theta_2B^2-\dots-\theta_qB^q)$, which is formed by taking all moving average terms of the ARIMA(p,d,q) model (i.e. right hand side of equation 3.4), when solved for B, must be greater than one.

3. Reasonableness of Forecasting. A model that can be fitted to a series adequately in the sense of passing the minimum level of significance in the goodness of fit test might nevertheless give unreasonable future projections, such as interest rates of 1000 percent. This is especially true when split-record tests are performed on a U-shaped or a W-shaped series. Therefore, a well-fitted model with a high level of significance is not necessarily a good projection model for the purpose of planning. A projection-reasonableness test is used as a supplementary criterion to the level of significance in selecting a model. A physical range and/or lower and upper bounds for the future are established based on available information and basic characteristics of the series. In this study, the bounds

are established at a distance of 50 times the range of values within the record from each extreme of record in terms of the transformed variable (logarithm, usually). A model is said to have projection reasonableness with respect to a series if projections of the series made by this model can satisfy all the designated range limitations. In this test, five independent projection sets, each with five times the record length, are generated randomly by the model. A model which cannot pass this test will be discarded.

4. Simplicity of Model. According to the principle of parsimony in using parameters, one should not try to overfit a time series with a complicated model when a simpler model can fit the series adequately. Thus, a trade-off between the performance (measured by the level of significance in the goodness of fit test) and the complexity (measured by the total number of parameters used) of models is desirable in selecting the best model for practical uses. The trade-off can be achieved by employing a discount factor, i , to the level of significance, α , such that α is reduced exponentially according to the complexity of the model. Mathematically, let $(p+q)$ be the degree of complexity of an ARIMA (p,d,q) , then the discounted level of significance, α' , is computed as

$$\alpha' = \alpha(1-i)^{p+q-1} \quad (4.9)$$

In selecting a value of i , personal judgment must be exercised. A value of 0.3 was used in this study, and this represents a high degree of discrimination against the more complex models.

Following the above four model screening steps, the best model can be selected simply by choosing the model which has the highest discounted significance level.

Applications of the ARIMA(p,d,q) Model. The ARIMA(p,d,q) model can be used to fit, to forecast, and to simulate a given time series.

1. Fitting a time series. Let \hat{W}_t be the fitted value of the observed time series W_t and a_t be the residual of fitting \hat{W}_t to W_t at time t . The first fitted value of \hat{W}_t to be computed using the fitted ARIMA model is \hat{W}_{p+1} , because the first p values of W_t are not used in the estimation

of autoregressive parameter ϕ_p . Thus for $t \leq p$, we have

$$\hat{W}_t = W_t \quad (4.5)$$

$$a_t = 0 \quad (4.6)$$

and for $t > p$, we have

$$\hat{W}_t = \phi_1 \hat{W}_{t-1} + \dots + \phi_p \hat{W}_{t-p} + \theta_0 - \theta_1 a_{t-1} - \dots - \theta_q a_{t-q} \quad (4.7)$$

$$a_t = \hat{W}_t - W_t, \quad (4.8)$$

and in equation (4.7), the random noise term, a_j with negative lag j is set to zero. It should be pointed out that W is the d th difference of the original observed series Z . Let \hat{Z}_t be the fitted value of the observed Z_t at time t , then the functional relationship of Z_t can be derived by substituting $W_t = \nabla^d Z_t$ into equation (4.7). Upon substituting and collecting similar terms, we have

$$\hat{Z}_t = \sum_{i=1}^{p+d} \psi_i \hat{Z}_{t-i} + \theta_0 - \sum_{i=1}^q \theta_i a_{t-i} \quad (4.9)$$

where ψ_i 's are coefficients of \hat{Z} . For example, if W is the first difference of Z_t (i.e. $W_t = \nabla Z_t = Z_t - Z_{t-1}$), then, by substituting $\hat{W}_t = \hat{Z}_t - \hat{Z}_{t-1}$ into equation (4.7), we have

$$\begin{aligned} \hat{Z}_t - \hat{Z}_{t-1} = & \phi_1 (\hat{Z}_{t-1} - \hat{Z}_{t-2}) + \phi_2 (\hat{Z}_{t-2} - \hat{Z}_{t-3}) + \dots + \\ & \phi_p (\hat{Z}_{t-p} - \hat{Z}_{t-p-1}) + \theta_0 - \theta_1 a_{t-1} - \dots - \theta_q a_{t-q}, \end{aligned} \quad (4.10)$$

collecting similar terms,

$$\begin{aligned} \hat{Z}_t = & (1-\phi_1) \hat{Z}_{t-1} + (\phi_2-\phi_1) \hat{Z}_{t-2} + \dots + (-\phi_p) \\ & \hat{Z}_{t-p-1} + \theta_0 - \theta_1 a_{t-1} - \dots - \theta_q a_{t-q} \end{aligned} \quad (4.11)$$

or, in terms of ψ_i ,

$$\hat{Z}_t = \sum_{i=1}^{p+1} \psi_i \hat{Z}_{t-i} + \theta_0 - \sum_{i=1}^q \theta_i a_{t-i} \quad (4.12)$$

2. Forecasting and Simulation. Let \hat{W}_{t+s} be the forecast of the series W_t at lead time s from the origin t . It can be shown (Box and Jenkins, 1970, Chapter 5) that: (a) the forecast for W_{t+s} using minimum mean square error estimation is the conditional expectation of W at time $t+s$; (b) the expected forecast error of \hat{W}_{t+s} is equal to zero, hence the forecast is unbiased; and (c) the one-step-ahead forecast error is equal to a_t (the residual at time t), hence the Chi-square significance test of a_t can be used as an effective indicator of the goodness of fit of the ARIMA model for the series being modeled. The forecasting equation for \hat{W}_{t+s} is

$$\hat{W}_{t+s} = \phi_1 \hat{W}_{t+s-1} + \dots + \phi_p \hat{W}_{t+s-p} + \theta_0^{-1} a_{t+s-1} - \dots - \theta_q a_{t+s-q} \quad (4.13)$$

$$\begin{aligned} \text{where } \hat{W}_t &= W_t, \text{ if } t \leq n-d, \\ &= \hat{W}_t, \text{ if } t > n-d, \\ a_t &= a_t, \text{ if } t \leq n-d, \\ &= E(a_t) = 0, \text{ if } t > n-d \text{ and } E(a_t) \text{ is the expected value} \\ &\quad \text{of the white noise } a_t. \end{aligned}$$

And similar to the derivation of the fitted value \hat{Z}_t from \hat{W}_t , the forecasting function for \hat{Z}_{t+s} can be shown to be

$$\begin{aligned} \hat{Z}_{t+s} &= \psi_1 \hat{Z}_{t+s-1} + \dots + \psi_{p+d} \hat{Z}_{t+s-p-d} + \theta_0^{-1} a_{t+s-1} \\ &\quad - \dots - \theta_q a_{t+s-q} \end{aligned} \quad (4.14)$$

$$\begin{aligned} \text{where } \hat{Z}_t &= Z_t, \text{ if } t \leq n \\ &= \hat{Z}_t, \text{ if } t > n, \text{ and} \\ a_t &= a_t, \text{ if } t \leq n \\ &= 0, \text{ if } t > n. \end{aligned}$$

The forecast \hat{Z}_{t+s} represents the most likely projection level for lead time s in the future. For reasons mentioned before, alternative projections are needed and can be generated by simulation using the fitted ARIMA model with random components.

Each independent set of simulated projections is generated using a

stochastic trend factor β and a white noise term b_t . The stochastic trend factor β is obtained by assuming that the trend factor, θ_0 , used in forecasting, has a Student-t distribution. The white noise term b_t is generated by a random variable generator using the known variance of the white noise a_t . That is

$$\begin{aligned} b_t &= a_t \text{ if } t \leq n \\ &= q \sqrt{\sigma_a^2}, \text{ if } t > n \end{aligned} \quad (4.15)$$

q is a normally distributed variable with zero mean and unit variance. Then, the simulation function of \hat{Z}_{t+s} is

$$\begin{aligned} \hat{Z}_{t+s} &= \psi_1 Z_{t+s-1} + \dots + \psi_{p+d} Z_{t+s-p-d} + \beta - \theta_1 b_{t+s-1} \\ &\quad \dots - \theta_q b_{t+s-q}, \end{aligned} \quad (4.16)$$

$$\begin{aligned} \text{where } \hat{Z}_t &= Z_t \text{ if } t \leq n \\ &= \hat{Z}_t \text{ if } t > n. \end{aligned}$$

5. MULTIVARIATE PROJECTION MODEL

General Approach. For variables that are interrelated, an acceptable model should project every variable simultaneously so as to preserve the interrelationship among projections of all variables. There is no specific multivariate projection technique, but principal-component analysis can be used where data for all variables exist for the same period, in order to form uncorrelated variables (principal components), each of which is expressed as a linear function of the original variables. Then the new variables can be projected independently and transformed to form projections of the original variables. The principal-component technique was selected for convenience in this study as an illustrative procedure. Other multivariate techniques, such as multiple linear regression, can be used if desired.

Principal-Component Analysis. Principal-Component analysis is a straight-forward method of transforming a given set of variables into a new set of composite variables (or principal components) that are mutually orthogonal (uncorrelated). The first principal component represents the single best summary of linear relationships exhibited in the data. The second component is defined as the second best linear combination of variables that accounts for the most residual variance after the effect of the first component is removed from the data. Subsequent components are defined similarly until all the variance in the data is exhausted. Unless at least one variable is perfectly determined by the remaining variables in the data, the principal-component solution requires as many components as there are variables. Let Z_j , $j=1, 2, \dots, m$ be the m original variables; F_k , $k=1, 2, \dots, m$ be the m principal components; A_{jk} , $j=1, 2, \dots, m$ and $k=1, 2, \dots, m$ be the component-pattern coefficients, and b_{jk} , $j=1, 2, \dots, m$ and $k=1, 2, \dots, m$ be the component-estimation coefficients. Then the principal component F_k can be computed from the values of the m original variables Z as

$$F_k = b_{k1} Z_1 + b_{k2} Z_2 + \dots + b_{km} Z_m, \quad k = 1, 2, \dots, m \quad (5.1)$$

and the original variable Z_j can be described by the m principal components F as

$$Z_j = A_{j1} F_1 + A_{j2} F_2 + \dots + A_{jm} F_m, \quad j = 1, 2, \dots, m. \quad (5.2)$$

It should be noted that a reduction in the number of principal components used in equation 5.2 is possible if a certain percentage of unexplained variance of the m original variables Z is allowed. For example, if the first m_1 ($m_1 < m$) principal components account for 98 percent of the variance (hence 2 percent unexplained variance) in the m original variables, and if a 2 percent unexplained variance in expressing the m original variables is allowed, then only the first m_1 principal components are needed in describing the m original variables. The value of m_1 is dependent on

the value of m and the maximum percentage of unexplained variance of the original variables allowed. The reduction in the number of principal components used from m to m_1 represents a way of reducing the size of the multivariate projection model, since only projections of m_1 principal components are needed to calculate projections of m original variables. In the demonstration examples herein, no reduction in variance was allowed, hence $m_1 = m$ in these cases.

Multivariate Projection Model. Simultaneous projection of a set of original variables can be obtained by the following steps using the principal-component analysis allied with the single-variable projection model, ARIMA(p,d,q):

1. Find the component-pattern coefficients, A_{jk} , and component-estimate coefficients, b_{jk} , by principal-component analysis. (For the purpose of this study, these coefficients were calculated by the computer program **BMD01M** which is written in FORTRAN and available from Biomedical Computer Programs (BMD) developed at UCLA for IBM system 360/370.)
2. Construct a time series for each of m principal components by substituting values of the m original variables in equation 5.1.
3. Apply the ARIMA(p,d,q) model as described in section 4 to project future values of the most likely projection level and of a number of simulation sets for each of m principal components independently.
4. Compute projections of the most likely projection level and simulations of the m original variables by substituting projections of principal components in equation 5.2.

6. TEST TECHNIQUE

Split-Record Testing. In order to develop some degree of confidence that projections using techniques studied herein would in fact represent the entire range of future possibilities, or at least to evaluate the degree to which they do represent the future range and possibly to develop correctional criteria, the split-record type of testing has been employed in this study. This consists of calibrating the projection models on the basis of one part of the record (usually half) and examining the range of projections in comparison with the actual variations that occurred in the remainder of the record. This technique has been particularly favored, because it conforms to the exact manner in which projections are employed in water resources planning. In the test procedure, the first half of the record might represent all of the record that would be available at the time of planning the project, and the second half of the record would represent what occurs during the operation of the project. Thus, in concept, the test would show how the various projections made in project planning would actually compare to events that really occurred during the project life.

The split-record technique should be evaluated on the basis of a substantial number of variables. It is possible that a particular technique might accidentally show exceptional results when applied to one or two variables, but it is not likely that consistently superior results will be accidentally obtained as the number of variables increases. Even when a substantial number of test variables are used, care must be exercised that a great many techniques are not arbitrarily tested and the best selected, because, if sufficient alternatives are tried, there is an increasing probability that one will accidentally show favorable results.

When records are studied in halves for the purpose of testing, it is often possible to make more efficient use of available data by reversing the halves and repeating the test procedure. Thus, a model calibrated on the first half could be used for projecting through the period of the second half, and a model calibrated in reverse on the second half could be used for projecting in reverse through the first half of the record. Depending on the nature of the trends, this double use of the data could actually

double the effectiveness of the test, but in cases where the trends are consistently upward or downward, this double use of the data might not contribute greatly to the test results insofar as they would apply to those particular variables. In order to maximize the similarity of the tests to actual application, halves were not reversed in this study.

It should be recognized that, even though the test procedures are limited to calibration of the model using only half of the data, actual projections for water resources planning studies should be based on models calibrated using all available data.

Sample Time Series. To facilitate the test, 31 time series of 17 economic, 10 population, and 4 hydrologic variables were collected and listed in table 1.

Selection of Representative Projections. In each split-record test, 51 stochastic projections were made. These projections are numbered in the order of magnitude of the final value at the end of the projection (end of the second half of the record). In order to illustrate the range of projections and, in fact, to represent the entire set of projections with a small number of projections, projection numbers 2, 8, 26, 44 and 50 were selected. These are illustrated in figures 1 to 37. In each figure, the heavy line represents the observed data, the line connecting the symbol, \oplus , represents the most likely projection level, and the other five plots represent the five selected representative projections. On the basis of the variables used in this study, it appears that variations can occur in the future with a substantial probability (in the order of once in 50 projections) that differ in magnitude from the extreme projections computed in this study by an amount exceeding the range of projections computed and illustrated herein. In order to take these "maverick" possibilities into account properly, it is proposed that the following weights be assigned to the five representative projections:

<u>Projection no.</u>	<u>Representing Projection</u>	<u>Probability</u>
2	1 - 3	.06
8	4 - 12	.18
26	13 - 39	.50
44	40 - 48	.18
50	49 - 51	.06
Total		<u>.98</u>

Two more projections, to be assigned .01 probability each, can be obtained by utilizing projections numbers 2 and 50 to obtain extreme projections A and B such that values for each year are defined as:

$$Z_A = 2Z_2 - Z_{50}$$

$$Z_b = Z_2 - 2Z_{50}$$

This gives 2 new (maverick) projections that differ from projections 2 and 50 by the difference between 2 and 50 values for each respective year in terms of the transformed variable, Z. The maverick projections would each have a weight (probability) of .01, thus completing the range of probability represented in the above tabulation. Figures 38 and 39 are used to demonstrate the use of these extreme projections for the variable Man-Hours in Farming and variable Construction Cost, respectively. In each chart, projection A is represented by a single line connecting the symbol, *, and projection B by a single bold line.

7. STUDY RESULTS

Analysis of Economic Variables. Of the 17 economic variables (Fig 1 to Fig 17), observed quantities at the end of the projection period were outside of the range of projection in 5 cases (Figures 6, 7, 8, 13, and 17). If the projection techniques adequately represented all future probabilities, only one of these would be expected to be outside the range (an expected frequency of 4 out of 52 random projections would lie outside the range of the second-largest and second-smallest of 51 projections).

A careful review of these projections will indicate that, while different types of models would have greatly improved the projections in individual cases, no systematic model will satisfactorily represent what actually happened in all cases. Likewise, once a person sees the outcome of an observed variable, it is often possible to explain the variations, usually in terms of related factors, but it should be remembered that this study is concerned only with statistical means of projecting a stochastic variable or a set of interrelated stochastic variables.

It becomes obvious, therefore, as has been generally recognized, that the probabilistic range of projections should include provision for possible radical changes not indicated by the model calibration data. This is partly due to the fact that no model simple enough to be calibrated satisfactorily can adequately represent the complex factors involved and partly due to the likelihood that factors of potentially major impact are not adequately reflected in a particular sample of data.

Analysis of Demographic Variables. Population projections were made individually for all 10 cases (Fig 18 to Fig 27), and then 6 cases (Fig 28 to Fig 33) were projected jointly. Models selected by the computer program are as follows:

<u>Principal-component No.</u>	<u>ARIMA(p,d,q) model</u>
1	0, 1, 1
2	1, 1, 0
3	1, 1, 0
4	0, 1, 1
5	2, 1, 0
6	0, 1, 1

In the joint projections, selection of extreme projections was based on the weighted total population at the end of the sequence, giving equal weight to each variable, and simultaneous projections for all variables were used. Thus, the second highest of 51 projections for the weighted total population was not necessarily the second highest projection for each variable. In fact, it would tend to be less extreme in most or all cases, thus reducing the range of projections for each variable below that obtained in the independent projections. Whether the projections should be made jointly or independently would depend on the application, and it is just as vital that they be made independently for independent applications as it is that they be made jointly for applications where their simultaneous impact is of primary concern.

Of the 10 independent projections, observed values at the end of the projection period lie outside the projection range in 2 cases, which is at least twice the expected frequency. Thus, again it is necessary to account for greater variation in potential projections than a reasonably simple model indicates.

Analysis of Streamflow Variables. In the cases of natural streamflows, projections (Fig 34 to Fig 37) appear to be reasonable in relation to

observed variations during the period of projection, except for the Columbia River, where much lower quantities occurred in some years than were projected to any year. There is a strong downward trend through the early years and until 1935, after which a period of 20 years showed a strong resurgence of streamflows. The model used for streamflows is based on steady-state projection, because this gives best results for streamflows in general. If the model using $d = 1$ (first differences projected were used), it is apparent that the flows from 1935 to 1955 could fall well above all projections. It is difficult to determine whether the selected model is inapplicable or the observed data represent rare occurrences. There is a tendency to jump to the conclusion that actual events that conform to a forecasted projection "prove" the projection to be acceptable and that when actual events do not conform, the projection was in error. This, of course, is an over-simplification.

It is pertinent to recognize that the streamflow model determined to be most suitable is a steady-state model ($d=0$), and that the more general and extensively tested monthly streamflow simulation model developed in the Hydrologic Engineering Center of the Corps of Engineers is more suited for generation of sequences of this type of variable.

8. APPLICATION

General. In discussing the application of techniques studied under this project, it must be remembered that the techniques are meant to be applied to basic variables that cannot be derived directly from other variables that are more easily predictable. That is, the extrapolation is strictly statistical, as it must be in the case of many variables.

In making such extrapolations, natural or physical bounds must be recognized. The natural lower asymptotic bound of zero is taken into account in this study by transforming the variable to a logarithm, which is then unbounded at the lower end. If a variable is asymptotically bounded only on the upper end, taking the logarithm of the difference between the upper limit and the variable would remove that bound. If there are both lower and upper asymptotic bounds, the following transform can be used:

$$Z = \log\left(\log \frac{B-A}{X-A}\right)$$

where B is the upper bound, A the lower bound, X is the observed variable, and Z is the transformed variable. Other transforms might be devised and perform better under specific circumstances, but this one will at least assure that impossible values are not predicted. The transformed variable is projected, and resulting values are transformed back to the original variable needed in the contemplated study.

When more than one variable are needed in the study and where these are interdependent, it is necessary to perform a principal-component analysis of the transformed variables and to project these components, then recombine them into the transformed variables before transforming back to the original variables.

All of the above provisions are incorporated into the computer program described herein.

Water Resource System Simulation. Because of the complexity of water resources planning and management studies, the most acceptable general procedure for analysis consists of postulating a system and operation rules, and simulating the operation of the system under anticipated system demands and system inputs. Traditionally, system demands have been projected as a most likely sequence of

demands, and system inputs have been assumed similar to past experience, adjusted for future changes in controlling conditions (such as urbanization and reservoir construction). It is the thesis of this study that a range of demands and inputs should be considered and that appropriate weights should be assigned to the results of different projections. The technique described above for computing 51 projections and selecting 5 of these plus 2 "maverick" projections is intended to represent the full range of future probabilities with a minimum of computational effort.

Computer programs such as HEC-3, "Reservoir System Analysis for Conservation", developed in the Hydrologic Engineering Center of the Corps of Engineers, are very useful in performing the detailed simulation studies necessary for evaluating system performance under each combination of projections of the pertinent input and demand variables. In the case of hydrologic variables such as streamflow, rainfall and evaporation, where the process can be treated as stationary and where stochastic variations from month to month are important, they should be projected using a model such as HEC-4, "Monthly Streamflow Simulation".

Integration of simulation outputs. To the extent that any of the input and demand variables are independent of each other, it will be necessary to give special consideration to the manner in which the different sequences of the various variables are combined. In extreme cases, it may be necessary to perform simulations corresponding to all combinations of sequences, but it is hoped that this will not be necessary in ordinary applications. For example, principal variations in physical results of a plan of operation may depend almost entirely on hydrologic sequences, in which case the detailed water resource simulation studies need be made only once for each hydrologic sequence. Evaluation of the results in terms of economic output would require integration of many combinations of these outputs with the various economic projections. The expected system output would then be the sum of the cross products of the output for each simulation times the probability associated with the particular input. The probability associated with any particular input when a combination of independent inputs

is used equals the product of the probabilities associated with the independent inputs. It must be assured that probabilities associated with all simulations add to 1.0.

Illustrative Projections. For illustration purposes, projections of sample variables to year 2020 were obtained by using the projection techniques and computer programs developed in this study. These are illustrated in figures 40 to 61. Variables which are considered to be interdependent are projected by use of the multivariate projection model. These are:

1. Per capita GNP, industrial production index, gross nonfarm product, and gross farm product (shown in figures 40, 41, 42, 43, respectively).
2. Output per unit of labor input index, and output per unit of capital input index (shown in figures 44, 45 respectively)
3. Wholesale price index and consumer price index (shown in figures 46 and 47 respectively).
4. Manhattan Island real estate mortgage rates, stock exchange call loan rates and commercial paper rates (shown in figures 48, 49, and 50, respectively).
5. Population of New England, Middle Atlantic, East North Central, West North Central, South Atlantic, and East South Central divisions (shown in figures 51, 52, 53, 54, 55, 56 respectively).

Variables projected by single-variable projection models are man-hours in manufacturing, man-hours in farming, number of operating business, unemployment rate, applications for patents, and construction cost index, and projections are shown in figures 57 to 62, respectively. It should be noted that projections illustrated herein are used only to demonstrate the projection techniques developed in this study and should not be used as a formal projection of those variables.

Computer Programs. Algorithms described herein are combined for convenient use into two computer programs -- one for single-variable projections, described in Appendix A, and one for multivariate projections, described in Appendix B.

9. CONCLUSION

Comparison of actual sequences of demographic and economic variables demonstrates that no projection technique can be depended upon to yield acceptable results if only a single sequence is projected. It is not possible to predict variations accurately, so it is essential that a range of projections be considered if errors of projection are important to a proposed study.

A model used for projecting such variables must be flexible enough to define variations reasonably, yet simple enough to be calibrated satisfactorily with available data. The two requirements are often unobtainable, and it becomes necessary in those cases to conform with the latter requirement. As a consequence, estimated errors of projection are usually smaller than true errors, and it has become necessary to expand the range of possible projections as described herein.

The autoregressive-moving-average family of projection models described herein provides a range of models from which one can be selected on the basis of goodness of fit to observed data and reasonableness of projection as described herein. The selected model can then be calibrated on the basis of all available data, and a set of projections generated to constitute input and demand data for any water resource system simulation study. Computer programs for accomplishing this are described in appendices A and B. The weighted average accomplishments of any proposed plan of development or management can then be computed as described herein, for the purpose of evaluating the plan under the complete range of future possible projections and computing and evaluation that should be far superior to any obtainable by conventional methodology based on single projections of each variable.

10. REFERENCES

A search of recent statistical literature concerning time series analysis especially in the area of forecasting and control was undertaken in order to compare methodologies and to apply newly advanced treatments to the problems addressed by this study. It was found that most of the related material published prior to 1970 was included in the book entitled Time Series Analysis; Forecasting and Control (Box and Jenkins, 1970). Chatfield and Pepper (1971) discussed some practical aspects of time-series analysis, Bhattacharyya (1974) employed the forecasting technique in time-series analysis to predict the demand for telephones in Australia, and Box and Jenkins (1974) presented some of the more recent advances in forecasting and control of a time-series.

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TABLE 1

RECORDED ECONOMIC, DEMOGRAPHIC AND HYDROLOGIC DATA

A11	PER CAPITA GNP (1958 DOLLARS, 1889=1966)	867	932	869	829	910	875	941	78	944
	805	847	1107	1141	1105	1164	1275	1271		1144
	1013	1022	1305	1331	1206	1224	1377	1328		1449
	1259	1246	1404	1562	1578	1593	1674	1667		1666
	1446	1403	1234	1208	1301	1413	1591	1664		1573
	1748	1565	2348	2679	3041	3126	2516	2408		2429
	1690	1818	2679	2771	2706	2855	2856	2847		2778
	2367	2519	3052	3125	3243	3393	3555			
	2899	2913								

A15	INDUSTRIAL PRODUCTION INDEX (1913=100, 1860=1959)	7,88	8,35	8,00	9,84	10,30	10,80	11,60	100	11,60
	7,48	7,49	14,40	13,90	13,50	13,40	15,50	17,50		17,50
	11,70	12,30	24,40	23,10	23,20	27,90	30,60	32,60		32,60
	20,30	22,30	34,70	33,70	39,70	36,90	44,70	49,20		49,20
	35,00	36,00	65,40	62,30	73,60	78,90	80,60	80,20		80,20
	50,60	56,70	100,00	94,10	109,30	129,60	129,70	113,20		113,20
	85,30	82,20	144,40	137,70	153,00	163,10	171,80	188,30		188,30
	124,00	100,10	119,90	129,70	149,10	178,30	194,50	188,00		188,00
	155,60	129,70	405,20	398,70	343,60	291,70	320,90	333,90		333,90
	213,90	275,50	447,30	421,40	473,20	489,40	492,70	457,00		457,00
	366,30	398,70								

A17	GROSS NONFARM PRODUCT (BILLIONS OF 1958 DOLLARS) 1889=1966)	38,8	44,4	41,9	39,8	45,5	43,3	47,9	78	48,3
	33,4	37,1	65,2	68,7	67,1	73,1	82,8	85,0		75,9
	54,3	55,9	94,9	101,8	90,4	92,0	110,3	103,6		110,4
	87,8	88,3	121,4	139,3	144,9	147,4	158,6	159,6		162,0
	115,5	116,9	103,3	100,7	114,3	126,2	148,2	156,3		144,5
	162,5	142,7	225,4	242,8	257,2	254,2	246,0	253,3		264,6
	160,2	177,1	321,4	338,0	332,4	361,7	368,5	374,0		367,8
	263,8	292,2	441,2	460,3	488,4	521,2	557,2			
	395,2	404,5								

RECORDED ECONOMIC, DEMOGRAPHIC AND HYDROLOGIC DATA

A21	GROSS	FARM PRODUCT	(BILLIONS OF 1958 DOLLARS)	1889-1966)	78	14.1	14.7
	11.7	12.0	11.4	11.5		15.8	16.1
	14.7	14.8	14.6	15.5		18.2	16.5
	15.7	15.3	18.1	17.5		18.3	17.9
	16.6	15.4	16.5	16.7		19.0	19.0
	18.4	17.4	19.7	15.6		18.6	20.5
	19.5	20.4	22.2	20.8		21.8	22.5
	19.9	19.9	20.6	21.9		21.8	22.5
	22.7	23.9	23.8	24.1			
A72	MAN-HOURS IN MANUFACTURING	(1958=100, 1889-1969)	81	39.6	40.8	40.8	
	35.7	38.1	40.8	38.7		66.1	56.6
	46.5	50.7	55.5	57.2		89.0	88.1
	64.4	67.8	70.2	70.4		73.2	72.4
	80.8	82.5	85.1	86.6		68.0	54.1
	76.9	65.4	42.6	48.1		102.5	102.2
	62.7	69.1	88.8	129.9		109.4	100.0
	92.9	99.6	107.7	114.1		125.7	128.2
	107.1	107.3	109.0	110.0			
	130.1						
A74	MAN-HOURS IN FARMING	(1958=100, 1889-1969)	81	196.0	197.8	197.8	
	181.3	185.5	187.2	189.0		209.8	211.2
	199.3	202.4	203.7	205.0		225.2	228.1
	212.5	213.8	218.2	221.0		217.5	221.5
	223.9	227.4	209.8	225.6		209.4	195.0
	219.5	217.2	222.0	191.7		166.7	162.2
	196.0	196.8	197.6	194.1		107.9	100.0
	162.7	150.0	141.9	125.3		69.6	68.0
	99.8	98.0	91.7	81.3			
	63.6						

TABLE 1

RECORDED ECONOMIC, DEMOGRAPHIC AND HYDROLOGIC DATA

A160	NUMBER OF OPERATING BUSINESSES (THOUSAND, 1870=1970)	558	603	639	637	101
427	494	558	603	639	637	661
747	864	905	920	970	994	1047
1111	1193	1114	1209	1152	1058	1106
1174	1281	1320	1357	1393	1418	1448
1515	1617	1655	1675	1708	1733	1708
1821	1996	2047	2113	2158	2172	2199
2183	1961	1974	1983	2010	2057	2116
2156	2023	1855	1909	2142	2405	2679
2687	2667	2632	2633	2629	2652	2708
2708	2544	2524	2527	2520	2519	2444
2442						

A163	OUTPUT PER UNIT OF LABOR INPUT INDEX (1958=100, 1889=1969)	28.9	29.6	30.9	30.1	81
26.8	30.0	28.9	29.6	30.9	30.1	32.1
32.6	33.1	33.7	34.0	34.3	36.6	36.4
37.2	37.3	38.4	36.3	37.6	39.4	37.2
43.0	45.4	47.2	49.7	49.5	50.5	51.1
53.5	53.2	52.1	56.2	57.8	59.2	59.3
63.5	65.7	65.8	70.1	74.0	71.7	71.0
76.8	83.8	87.1	90.3	94.4	94.4	96.4
103.0	112.4	115.6	119.4	122.9	125.4	127.4
132.0						131.1

A165	OUTPUT PER UNIT OF CAPITAL INPUT INDEX (1958=100, 1889=1969)	55.7	52.4	57.0	53.6	81
58.0	61.7	55.7	52.4	57.0	53.6	57.6
60.4	62.5	63.3	60.4	63.3	68.3	66.8
65.3	66.0	66.6	59.4	59.7	67.8	64.0
67.3	66.4	74.3	73.9	74.7	76.9	75.6
77.5	56.2	56.8	62.1	69.8	78.6	81.6
83.9	102.9	108.6	115.7	115.6	109.4	105.6
102.4	106.7	108.0	103.2	108.4	105.6	103.4
104.5	108.0	109.4	111.9	114.5	115.3	112.7
112.3						113.7

TABLE 1

RECORDED ECONOMIC, DEMOGRAPHIC AND HYDROLOGIC DATA

B1	UNEMPLOYMENT RATE (1890-1971)	18.4	13.7	14.4	14.5	82
4.00	3.00	11.7	13.7	14.4	14.5	12.4
5.00	3.70	3.90	4.30	1.70	2.80	8.00
5.90	4.60	4.30	8.50	5.10	4.60	1.40
5.20	6.70	2.40	3.20	1.80	3.30	4.20
8.70	11.7	24.9	20.1	16.9	14.3	19.0
14.6	15.9	24.9	1.90	3.90	3.90	3.80
5.60	9.90	4.70	1.90	4.10	4.30	6.80
5.50	3.30	3.00	4.40	3.80	3.80	3.60
4.90	6.70	5.50	4.50	3.80	3.80	6.50
	5.90	5.20	4.50	3.80	3.80	5.10
						1.40
						3.20
						17.2
						5.90
						5.50
						3.50

B50	APPLICATIONS FOR PATENTS (1860-1970)	10664	15269	21276	20420	111
7653	4643	6014	6932	21276	20420	19271
19171	19472	20414	21602	20308	20260	20059
22395	25556	34311	35422	35461	35684	40464
40930	40443	38353	38344	47811	35758	41337
41898	46334	50059	51986	58575	61273	65642
64448	68904	70177	70228	70135	59581	80337
86575	93063	80333	80622	91692	92364	94272
93752	83967	60185	61070	72576	75006	71306
69484	59609	49827	59295	83179	75847	74660
74108	64788	69631	82745	79012	82552	83587
84246	87921	90077	92976	92719	98402	103993
109052		90837				
		100150				
		83070				
		75964				
		63917				
		84290				
		69872				
		54815				
		40608				
		35559				
		21638				
		10664				
		15269				
		21425				
		35806				
		43905				
		56277				
		70759				
		85708				
		69143				
		91826				
		79834				
		93250				
		92719				

RECORDED ECONOMIC, DEMOGRAPHIC AND HYDROLOGIC DATA

MANHATTAN ISLAND REAL ESTATE MORTGAGE RATES (1879-1970)	92
5.92	5.78
5.13	5.45
5.05	5.17
5.35	5.35
5.65	5.97
5.92	5.95
5.05	5.03
4.93	4.95
5.71	5.85
7.06	7.73
5.43	5.65
5.20	5.20
5.18	5.09
5.50	5.46
5.91	5.95
5.60	5.77
4.77	4.98
5.09	5.03
5.89	5.90
5.35	5.65
5.16	5.20
5.35	5.18
5.58	5.46
5.92	5.95
5.45	5.60
4.71	4.77
5.15	5.09
5.88	5.89
5.20	5.65
5.04	5.20
5.68	5.18
5.50	5.46
5.89	5.95
5.09	5.77
4.74	4.98
5.19	5.03
6.04	5.90
5.14	5.78
5.04	5.45
5.45	5.17
5.47	5.35
5.88	5.97
5.11	5.95
4.80	5.03
5.42	4.95
6.18	5.85
5.20	7.73
4.96	
5.60	
5.55	
5.85	
5.00	
4.91	
5.58	
6.42	

STOCK EXCHANGE CALL LOAN RATES (1860-1970)	111
5.99	5.76
5.72	5.56
4.86	5.76
5.84	3.42
2.94	4.00
2.98	2.57
7.74	5.97
2.94	1.74
1.00	1.00
1.63	2.17
4.99	4.50
7.95	7.95
6.17	6.19
3.11	14.24
1.66	3.71
1.88	4.57
4.44	3.71
1.92	3.22
4.18	4.86
0.56	1.16
1.00	1.00
3.20	3.06
4.69	4.50
6.17	6.19
3.11	14.24
1.66	3.71
1.88	4.57
4.44	3.71
1.92	3.22
4.18	4.86
0.56	1.16
1.00	1.00
3.20	3.06
4.69	4.50
5.23	5.23
3.35	3.35
4.03	4.03
4.28	4.28
6.54	6.54
2.62	2.62
4.50	4.50
0.91	0.91
1.16	1.16
4.03	4.03
5.78	5.78
7.54	7.54
4.22	4.22
2.51	2.51
2.18	2.18
1.92	1.92
5.28	5.28
6.04	6.04
1.00	1.00
1.63	1.63
3.72	3.72
6.33	6.33
10.29	10.29
5.44	5.44
4.48	4.48
5.08	5.08
2.71	2.71
6.32	6.32
7.61	7.61
1.00	1.00
1.63	1.63
4.22	4.22
7.96	7.96

COMMERCIAL PAPER RATES (1890-1970)	81
6.91	6.48
5.71	5.40
5.72	5.81
7.50	4.75
3.59	6.62
0.56	2.64
1.45	0.54
3.85	2.16
7.72	2.97
5.80	5.40
5.18	5.81
4.01	4.75
4.02	4.52
0.76	2.73
0.75	0.66
2.18	2.33
4.38	3.26
5.80	5.40
5.18	5.81
4.01	4.75
4.02	4.52
0.76	2.73
0.75	0.66
2.18	2.33
4.38	3.26
7.02	6.48
6.25	5.40
3.84	5.81
4.34	4.75
0.75	6.62
0.81	2.64
1.44	0.54
2.46	2.16
5.90	2.97
4.72	5.40
6.66	5.81
5.07	4.75
4.11	4.52
0.94	2.73
1.03	0.66
3.81	2.33
5.10	3.26
5.34	5.40
5.00	5.81
6.02	4.75
4.85	4.52
0.81	2.64
1.44	0.54
2.46	2.16
5.90	2.97
5.50	5.40
4.67	5.81
5.37	4.75
5.85	4.52
0.59	2.64
1.49	0.54
3.97	2.16
7.83	2.97

TABLE 1

RECORDED ECONOMIC, DEMOGRAPHIC AND HYDROLOGIC DATA

MEAN ANNUAL FLOW=ST. LAWRENCE R. (CFS, 1861=1965)		105	
275000	284000	273000	266000
259000	224000	237000	231000
232000	256000	251000	253000
259000	229000	242000	216000
227000	228000	242000	238000
219000	229000	258000	241000
234000	229000	217000	221000
218000	218000	208000	183000
213000	214000	248000	217000
263000	276000	254000	267000
229900	210300	205400	200800
			191800
		272000	242000
		252000	252000
		269000	241000
		218000	228000
		244000	262000
		241000	247000
		251000	251000
		255000	263000
		216000	218000
		241000	258000
		223000	222000

MEAN ANNUAL FLOW=MISSISSIPPI R. (CFS, 1862=1965)		104	
201300	124000	87950	162300
136700	178600	126700	186800
305100	236000	201000	239100
263500	185200	111700	81500
152600	261100	224200	181800
207700	145900	121100	256000
200700	132800	186900	150200
159800	136600	67700	187600
236200	234700	209100	223100
232200	141100	113500	130300
218900	113200	99560	198200
		124700	206700
		287800	137300
		218700	120600
		157200	171000
		226000	204000
		125000	171900
		204100	264400
		156800	148000
		162900	169800
		145100	136000
		113000	89000
		66000	146000
		93000	70000
		29750	80010
		95920	109700
		40820	47200
		47450	80110
		113000	86000
		66000	119000
		93000	69000
		29750	41090
		95920	61950
		40820	35060
		47450	59850

MEAN ANNUAL FLOW=MISSOURI R. (CFS, 1898=1970)		73	
78000	92000	64000	77000
107000	104000	82000	51000
67000	77000	97000	82000
99000	114000	55400	37100
56640	53170	31020	47360
79920	92740	92210	139400
73520	57100	79170	79180
66110	107500	84190	84920
		116000	113000
		69000	66000
		78000	93000
		51900	29750
		96540	95920
		55300	40820
		44980	47450
		116000	89000
		69000	146000
		78000	70000
		51900	80010
		96540	109700
		55300	47200
		44980	80110
		116000	86000
		69000	119000
		78000	69000
		51900	41090
		96540	61950
		55300	35060
		44980	59850

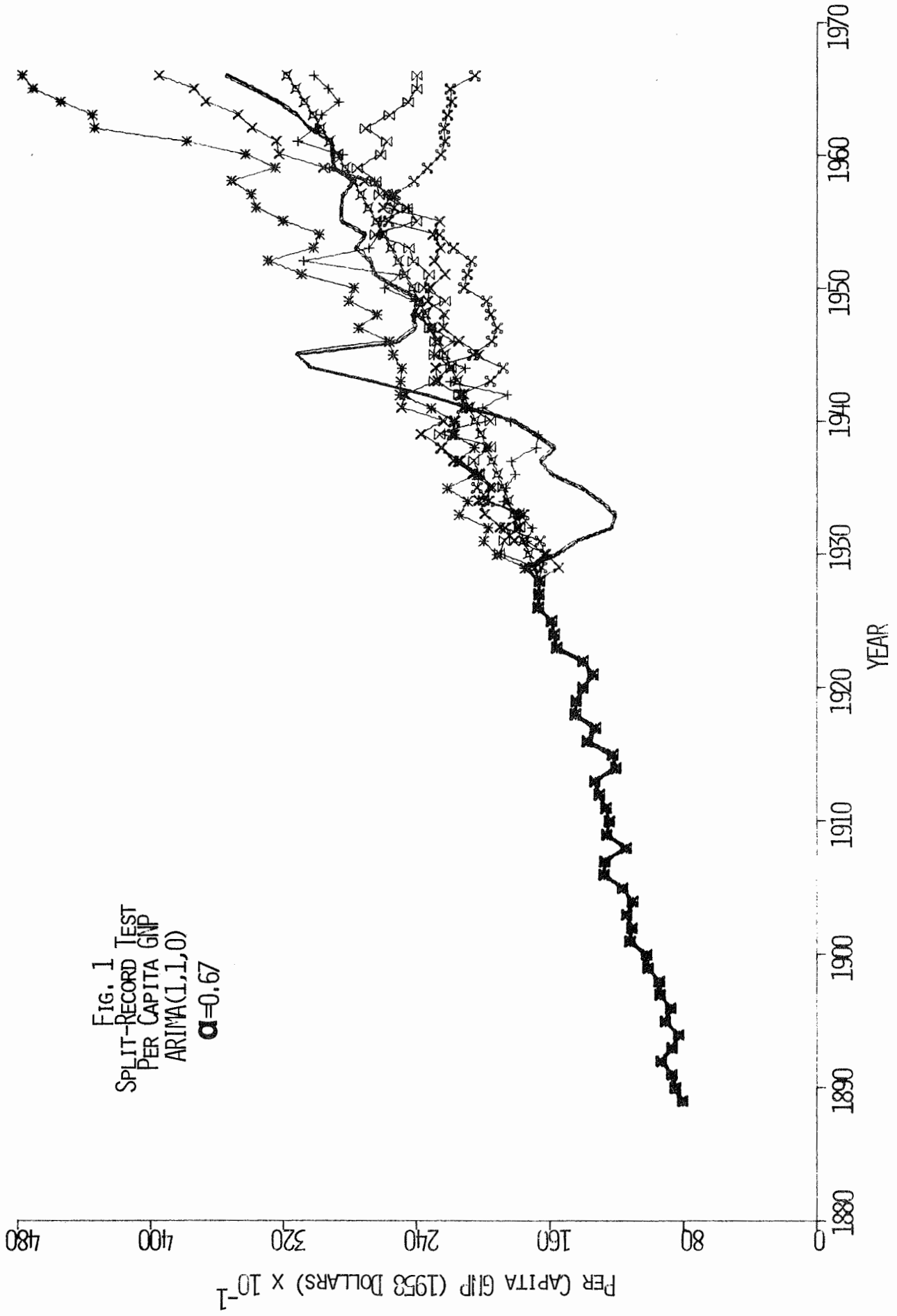


FIG. 1 TEST
 SPLIT-RECORD
 PER CAPITA GNP
 ARIMA(1,1,0)
 $\alpha=0.67$

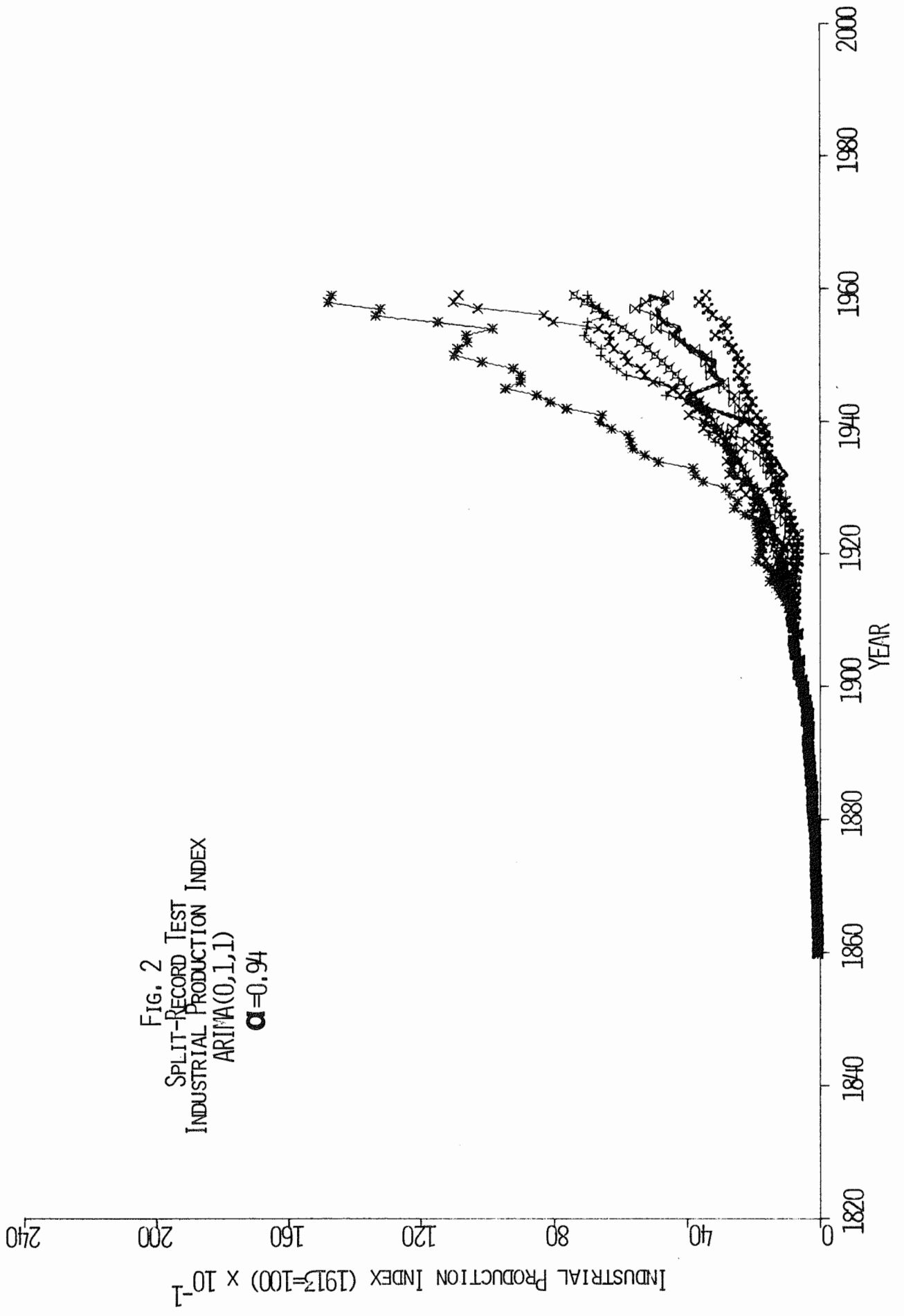


FIG. 2
 SPLIT-RECORD TEST
 INDUSTRIAL PRODUCTION INDEX
 ARIMA(0,1,1)
 $\alpha = 0.94$

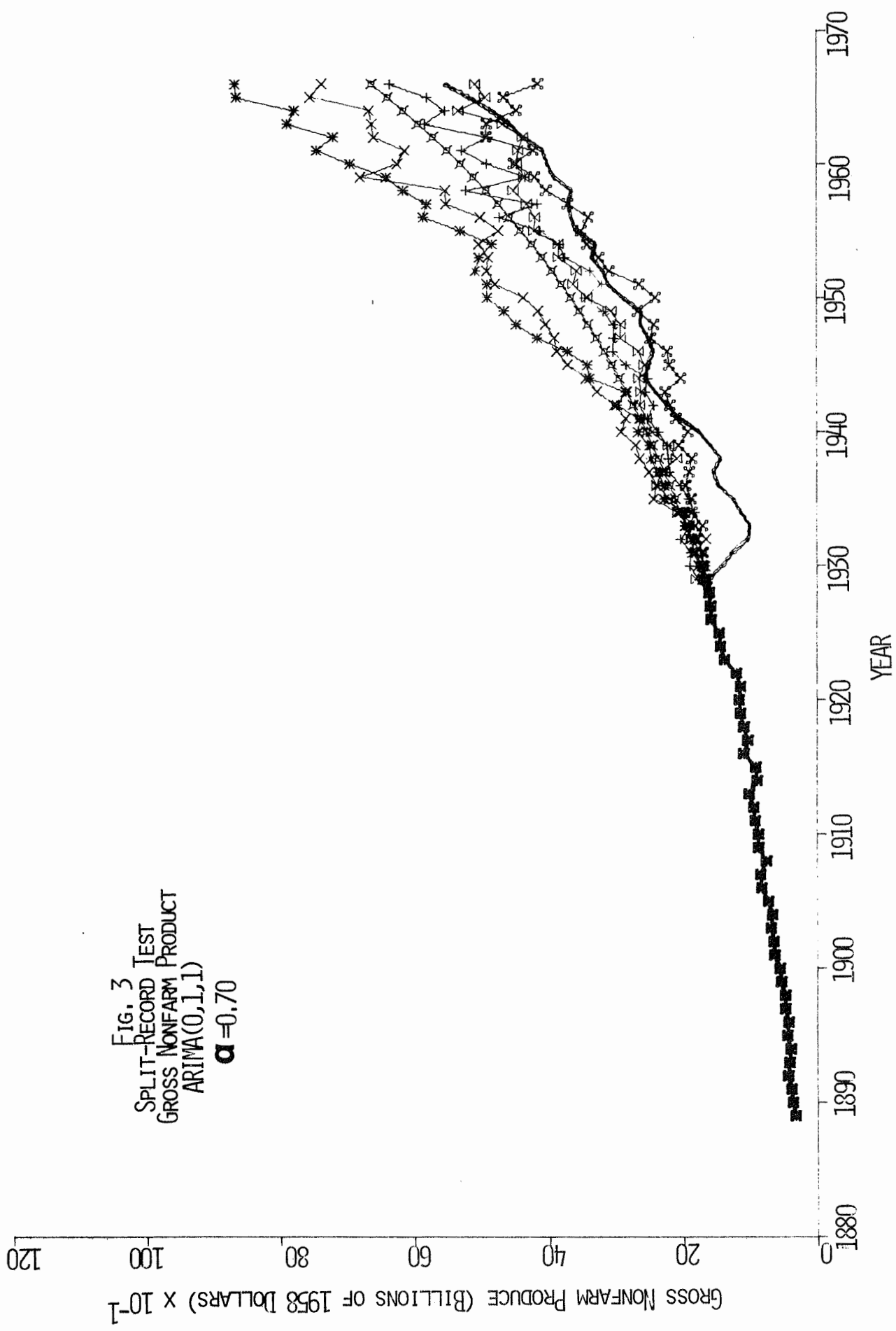


Fig. 3
 SPLIT-RECORD TEST
 GROSS NONFARM PRODUCT
 ARIMA(0,1,1)
 $\alpha = 0.70$

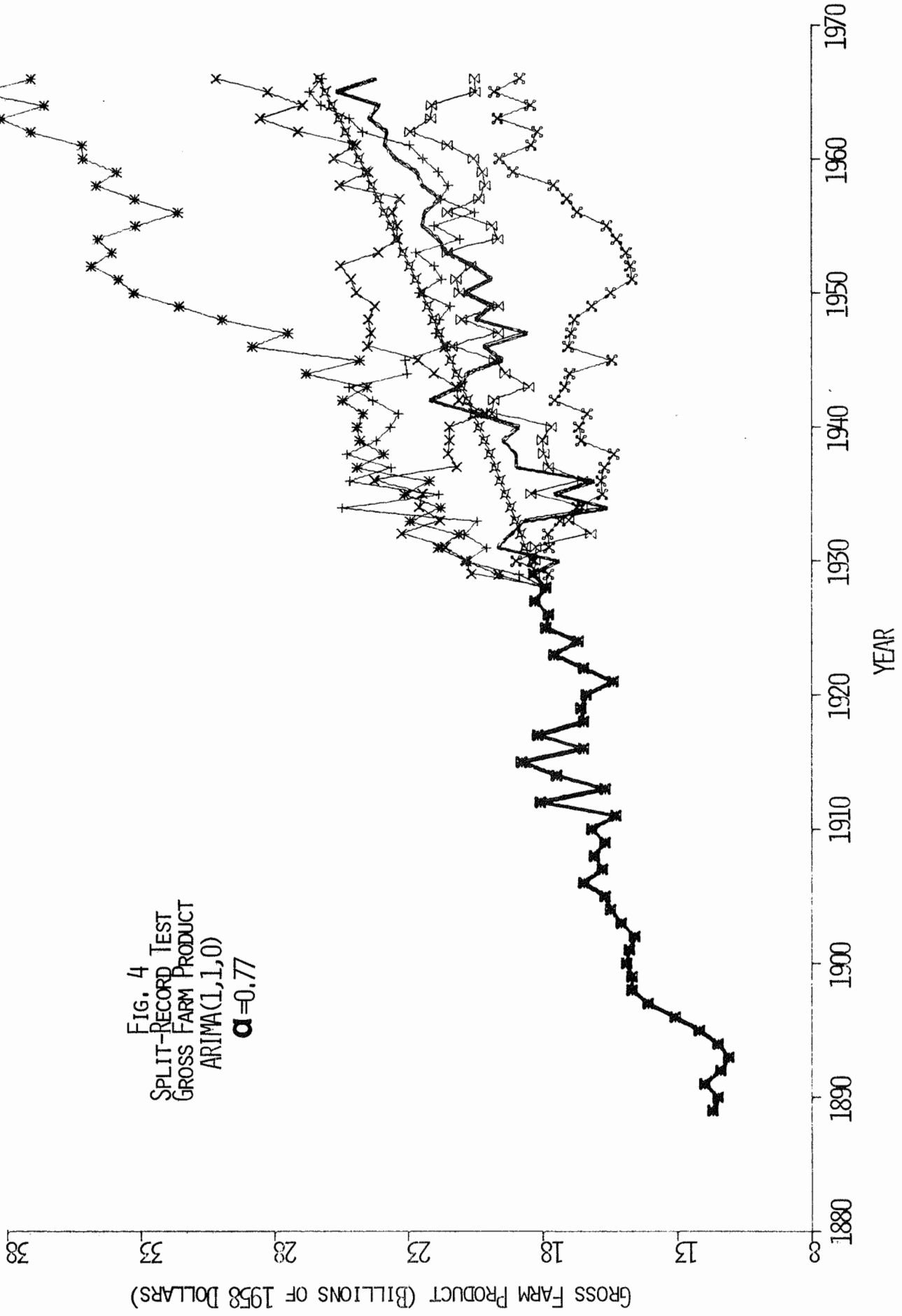


FIG. 4 TEST
 SPLIT-RECORD TEST
 GROSS FARM PRODUCT
 ARIMA(1,1,0)
 $\alpha = 0.77$

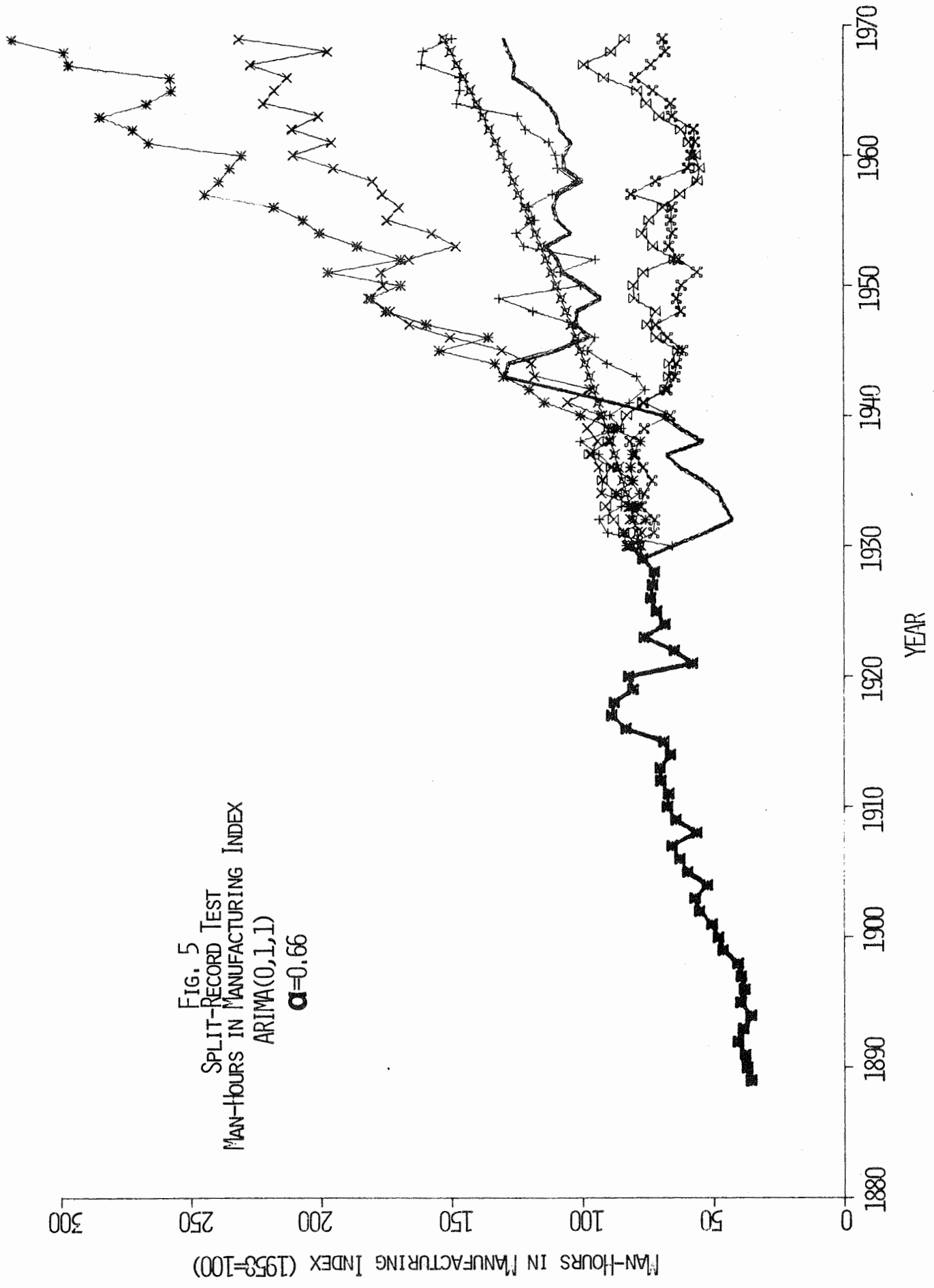


FIG. 5
 SPLIT-RECORD TEST
 MAN-HOURS IN MANUFACTURING INDEX
 ARIMA(0,1,1)
 $\alpha=0.66$

Man-Hours in Manufacturing Index (1953=100)

YEAR

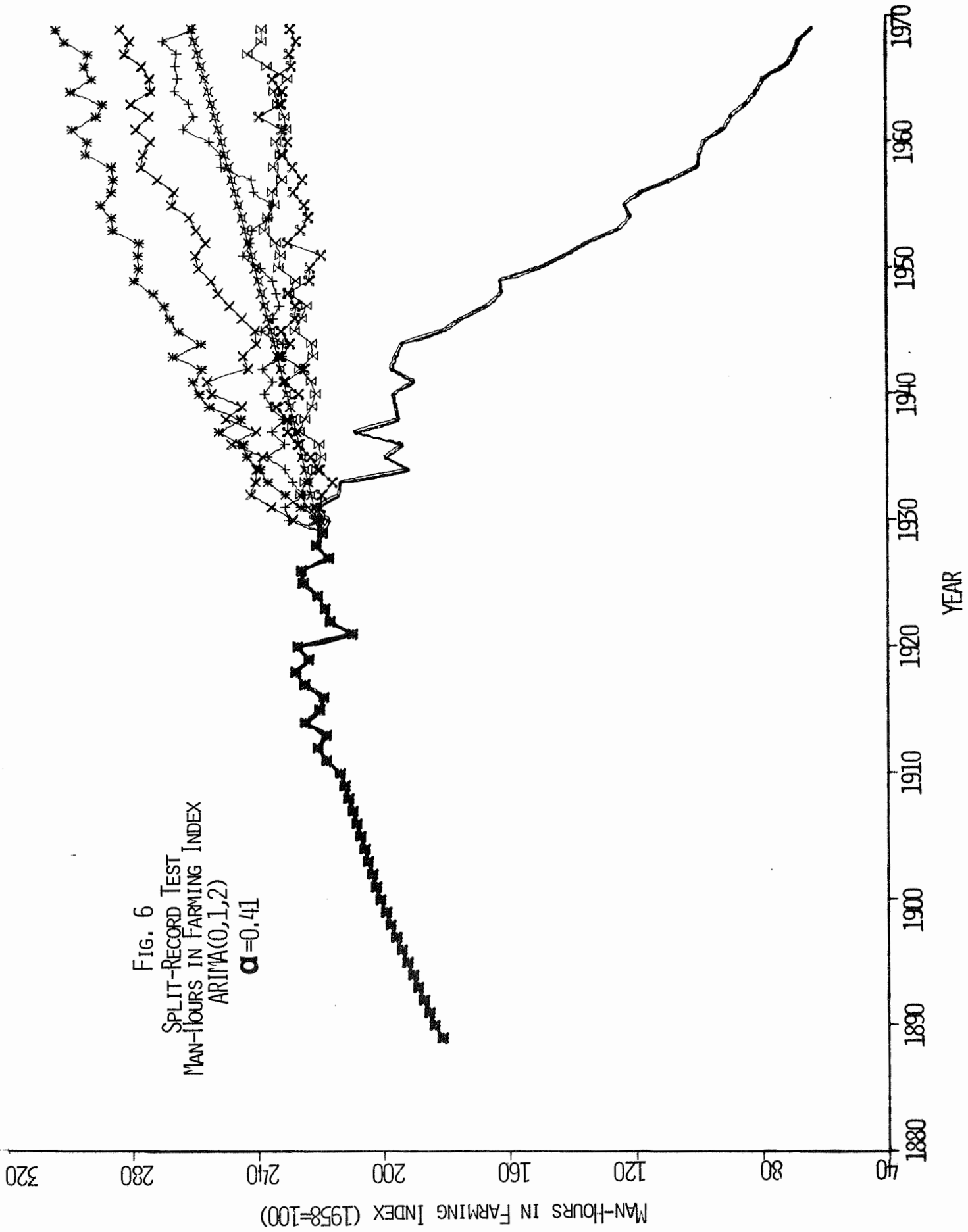


FIG. 6
 SPLIT-RECORD TEST
 MAN-HOURS IN FARMING INDEX
 $ARI(1,1,2)$
 $\alpha = 0.41$

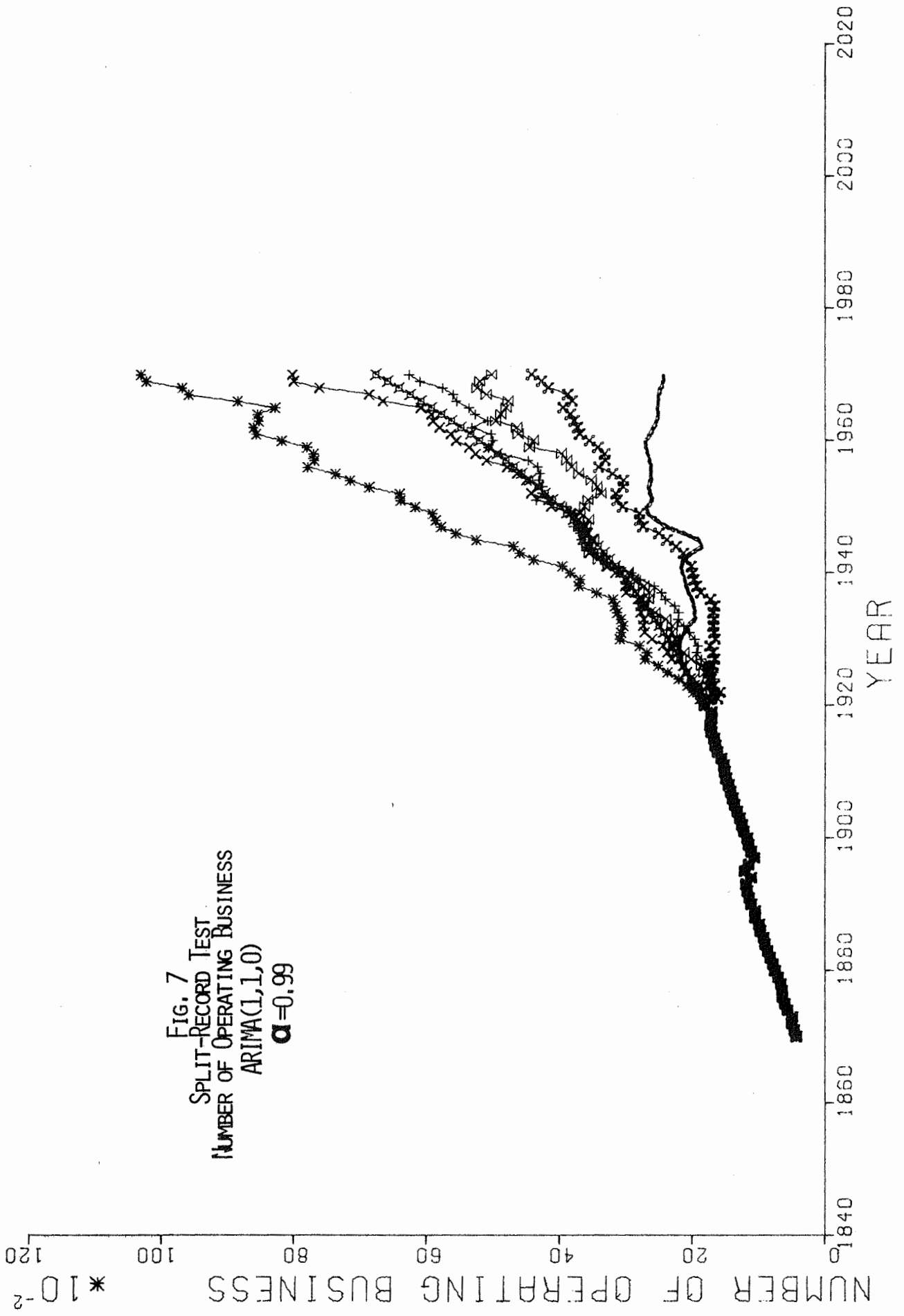


FIG. 7
 SPLIT-RECORD TEST
 NUMBER OF OPERATING BUSINESS
 ARIMA(1,1,0)
 $\alpha=0.99$

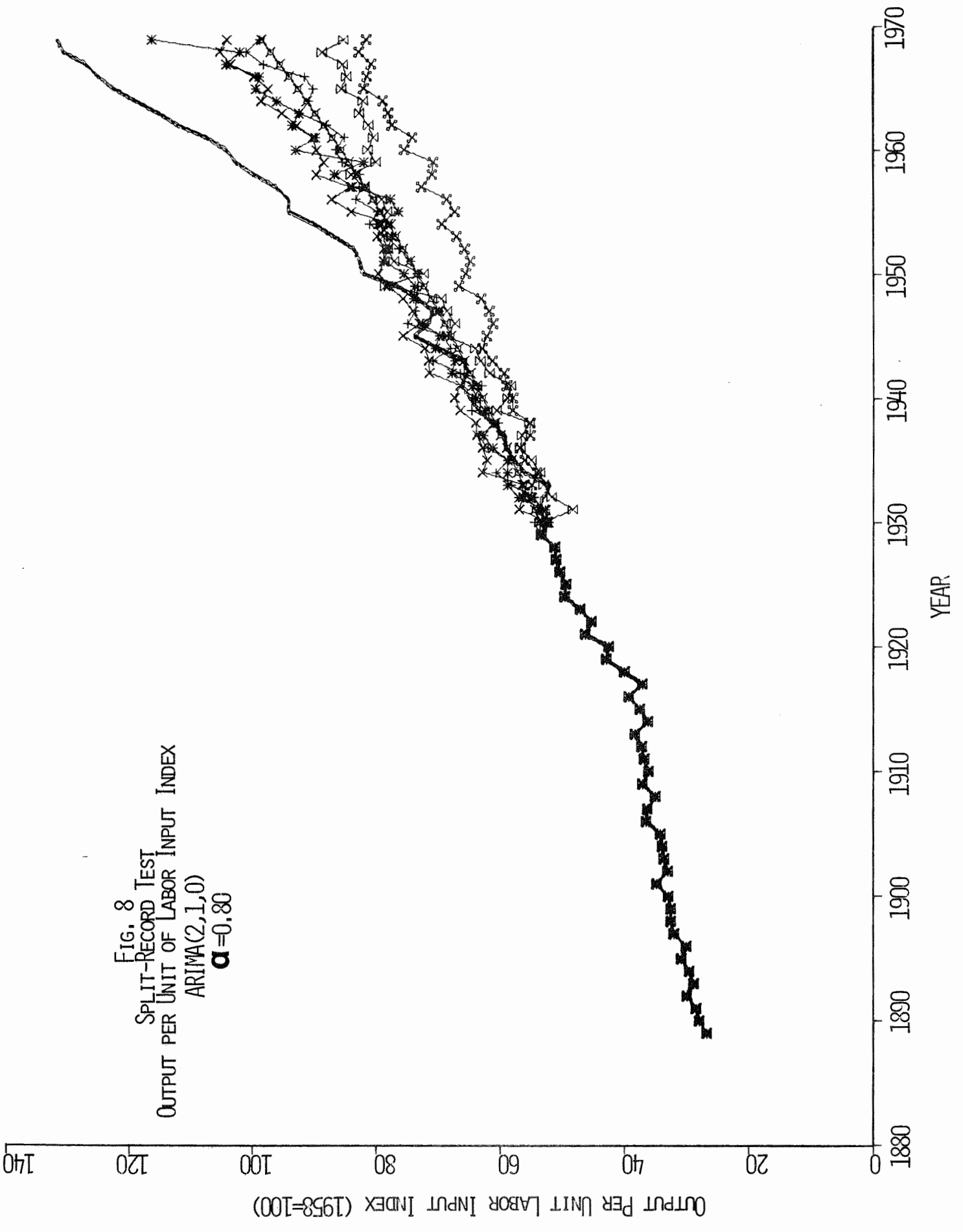


FIG. 8
SPLIT-RECORD TEST
OUTPUT PER UNIT OF LABOR INPUT INDEX
ARIMA(2,1,0)
 $\alpha = 0.80$

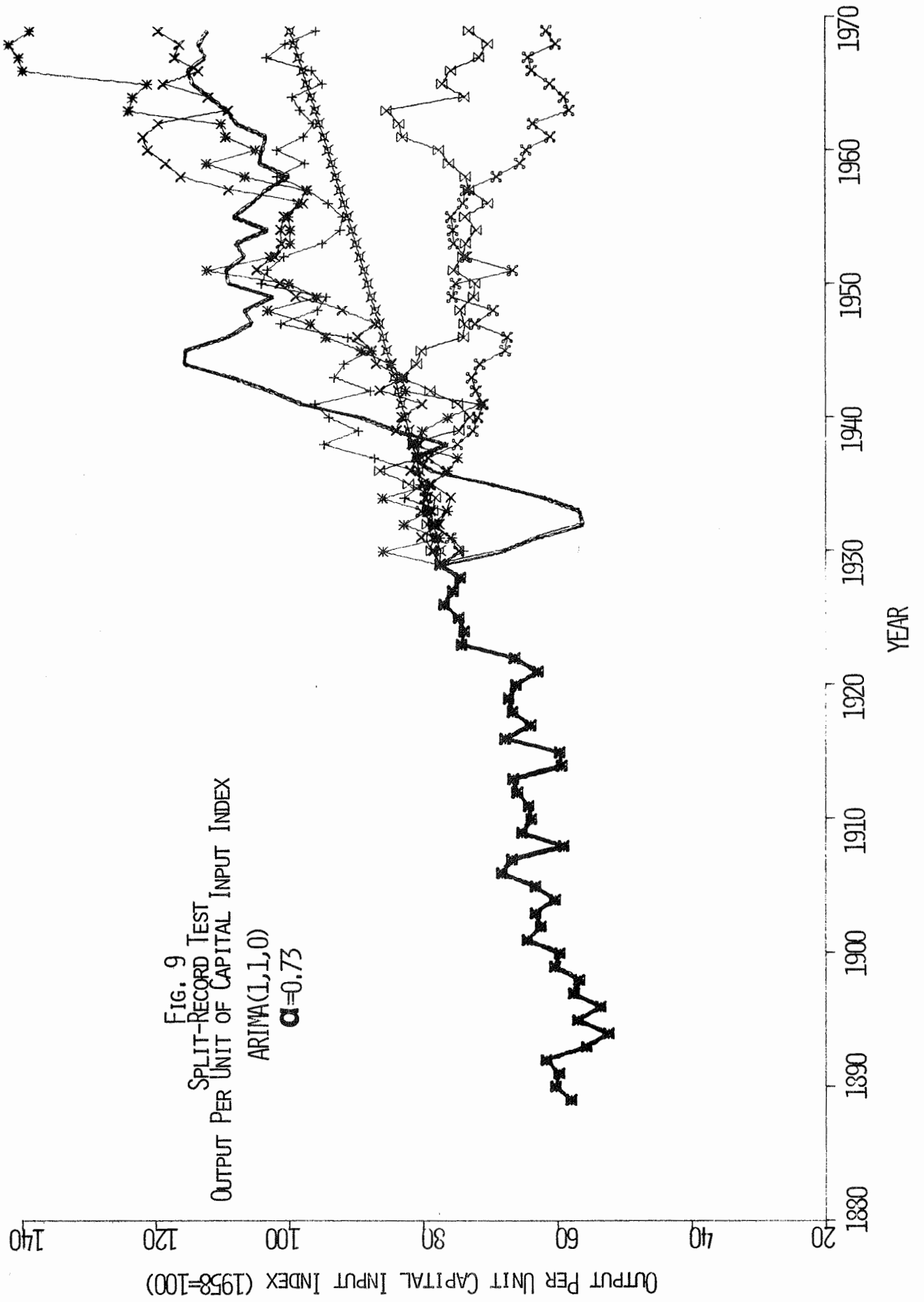


FIG. 9
 SPLIT-RECORD TEST
 OUTPUT PER UNIT OF CAPITAL INPUT INDEX
 ARIMA(1,1,0)
 $\alpha=0.73$

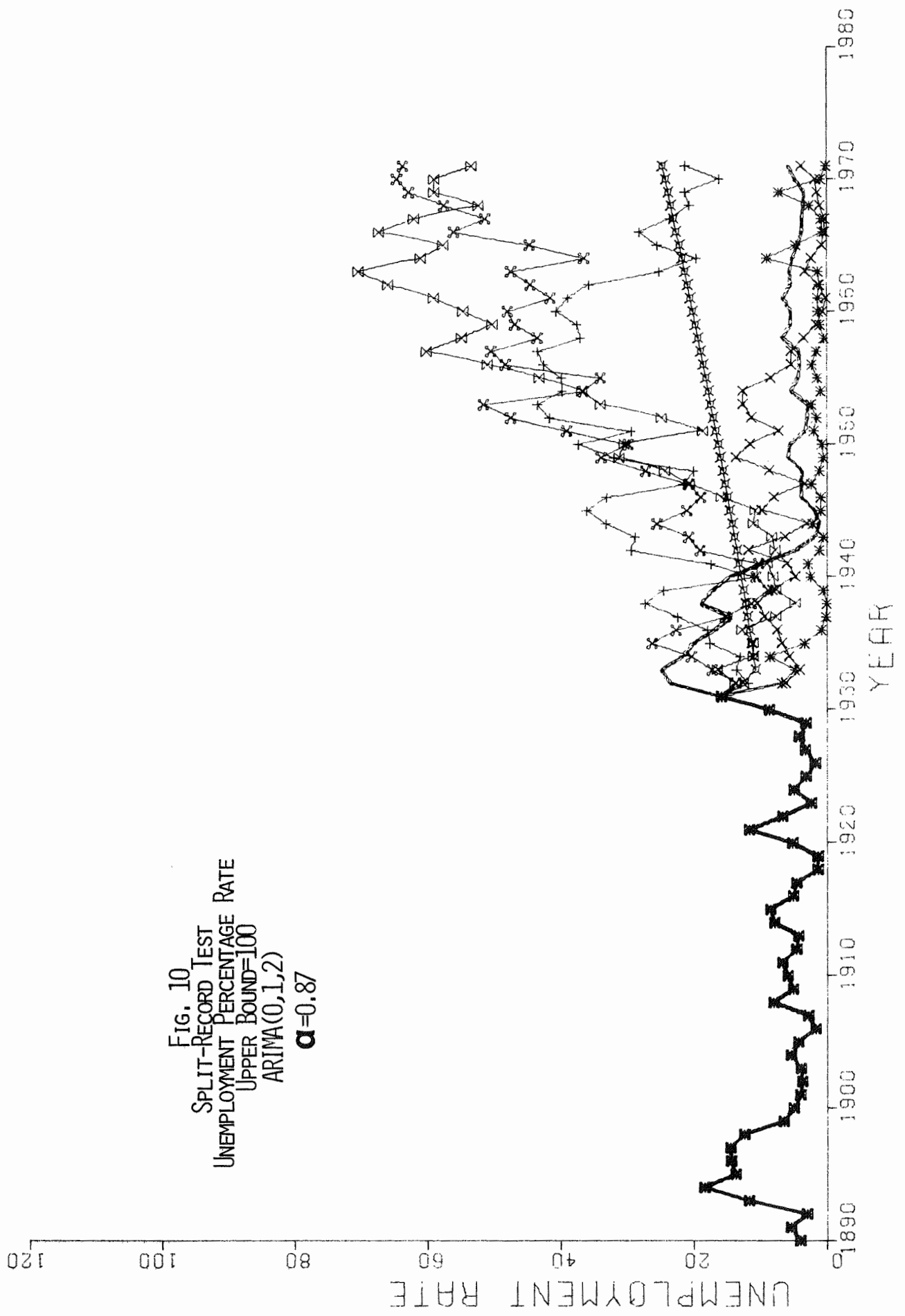


FIG. 10
 SPLIT-RECORD TEST
 UNEMPLOYMENT PERCENTAGE RATE
 UPPER BOUND=100
 ARIMA(0,1,2)
 $\alpha=0.87$

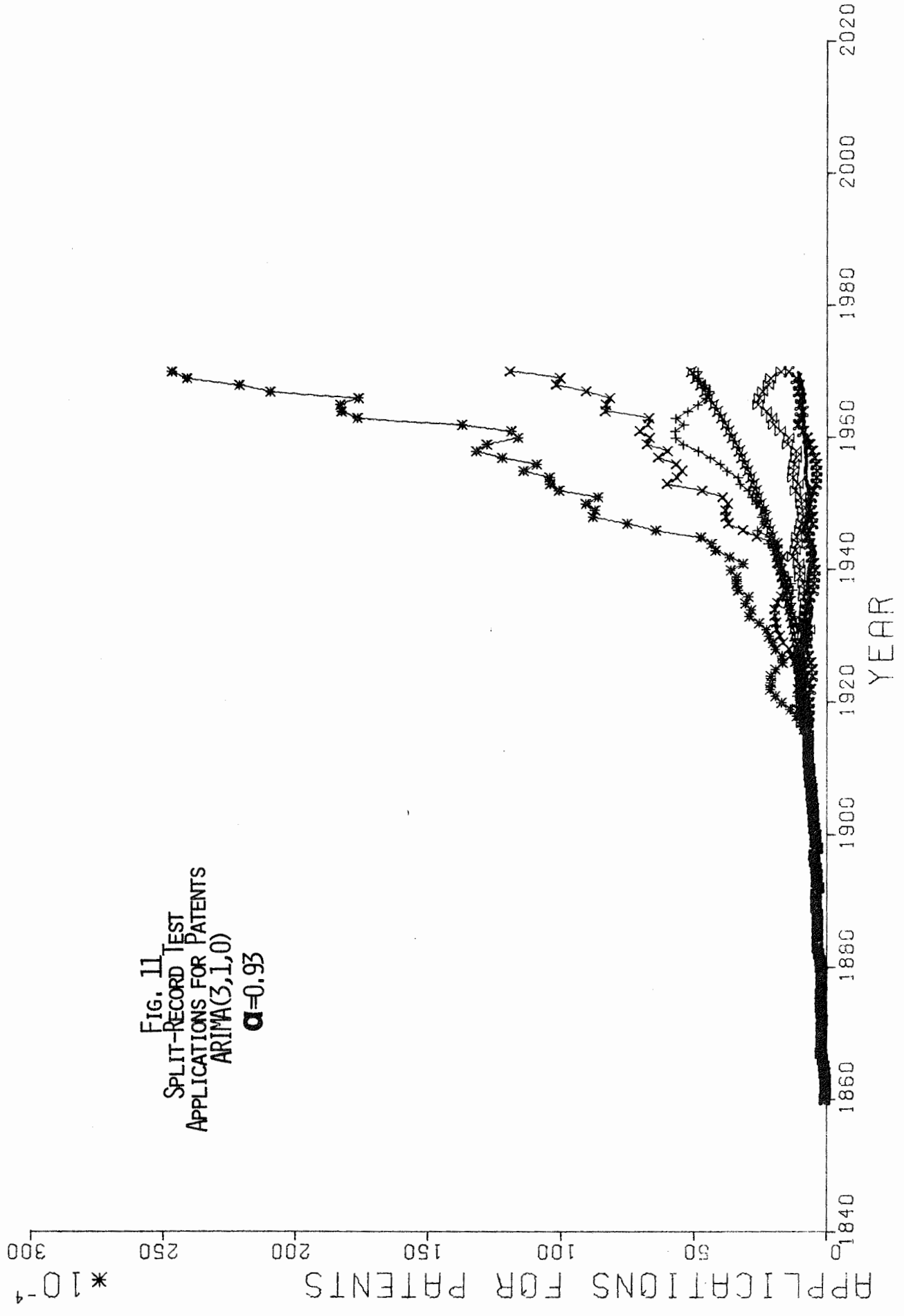


Fig. 11
 SPLIT-RECORD TEST
 APPLICATIONS FOR PATENTS
 ARIMA(3,1,0)
 $\alpha=0.93$

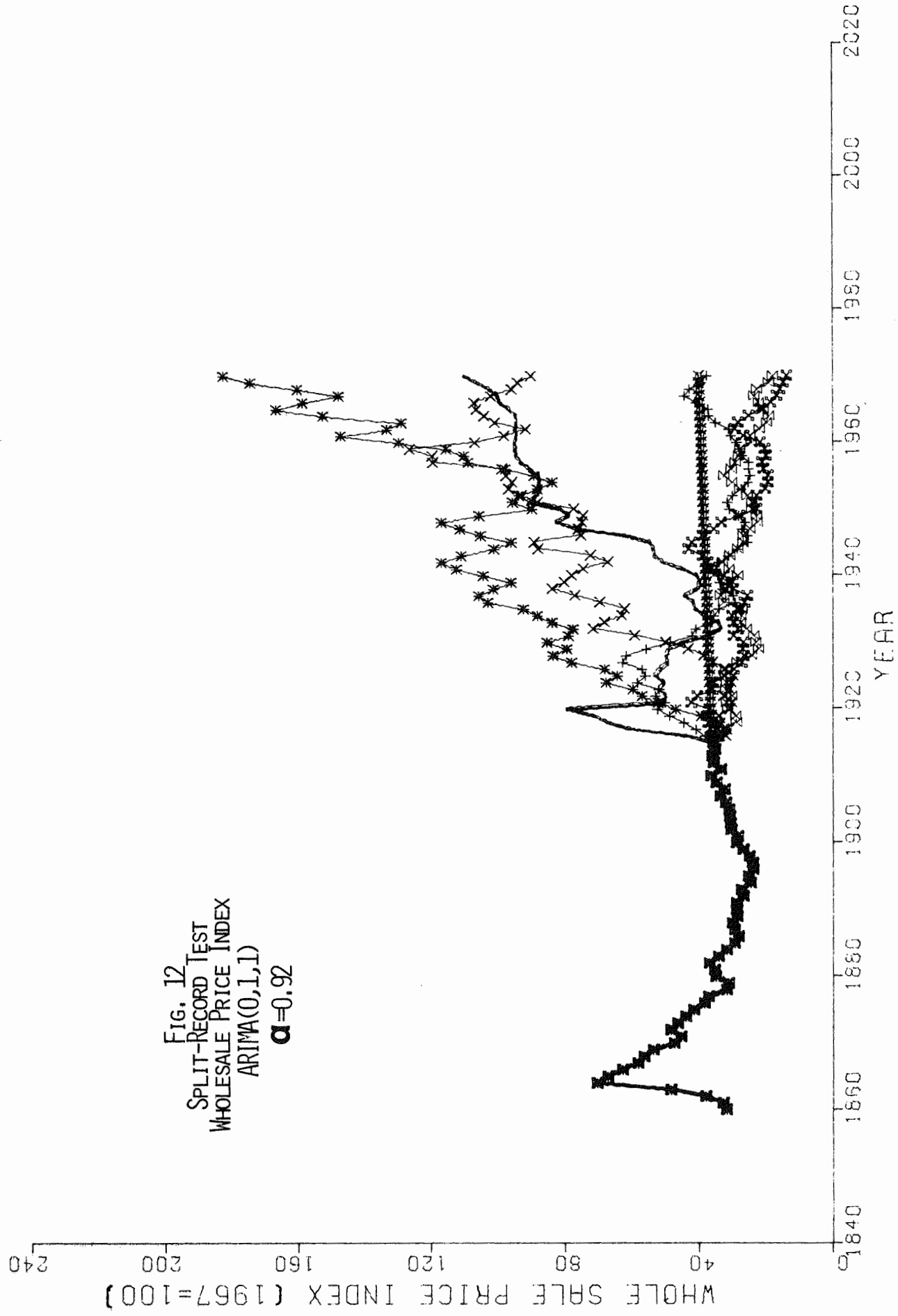
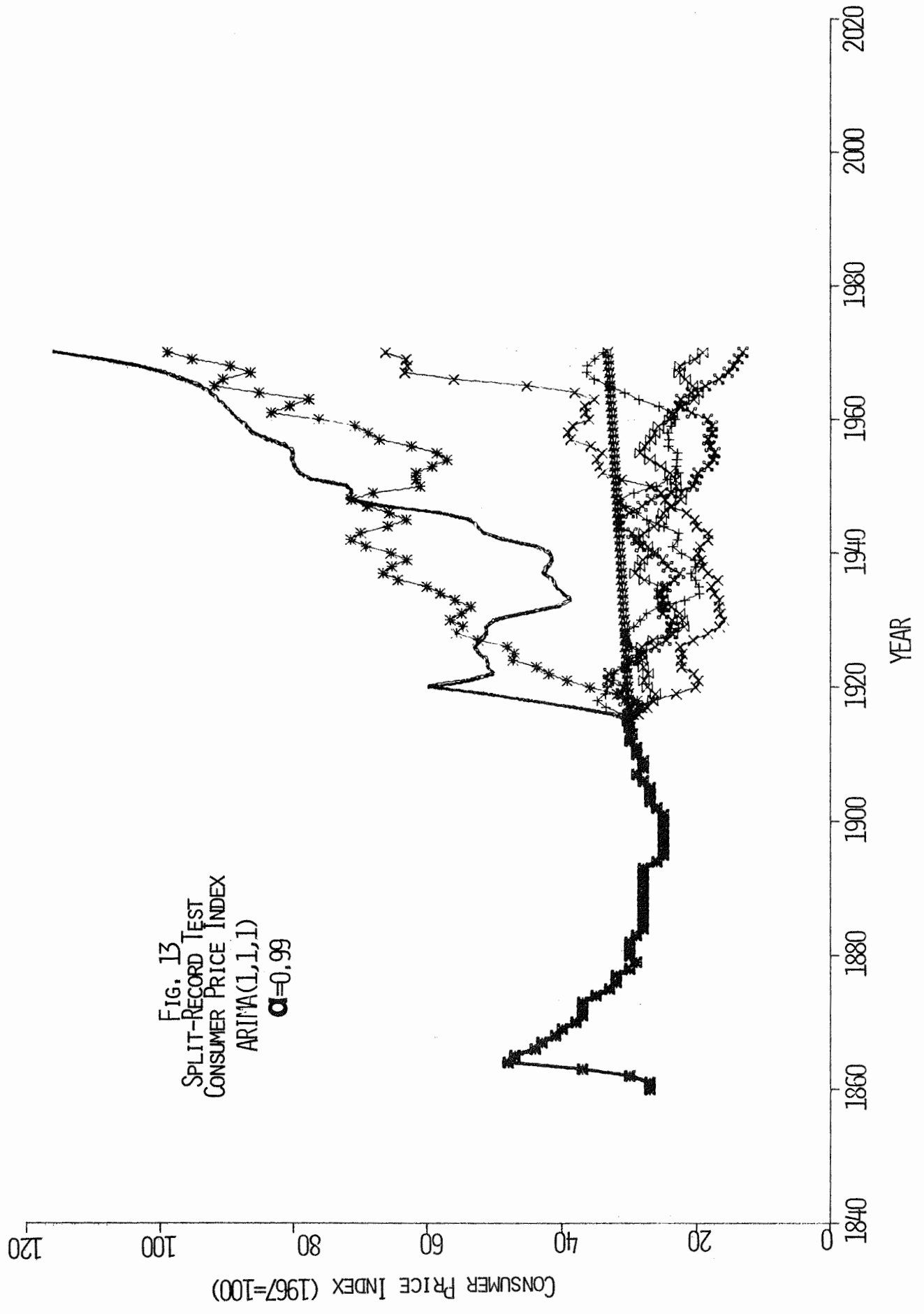
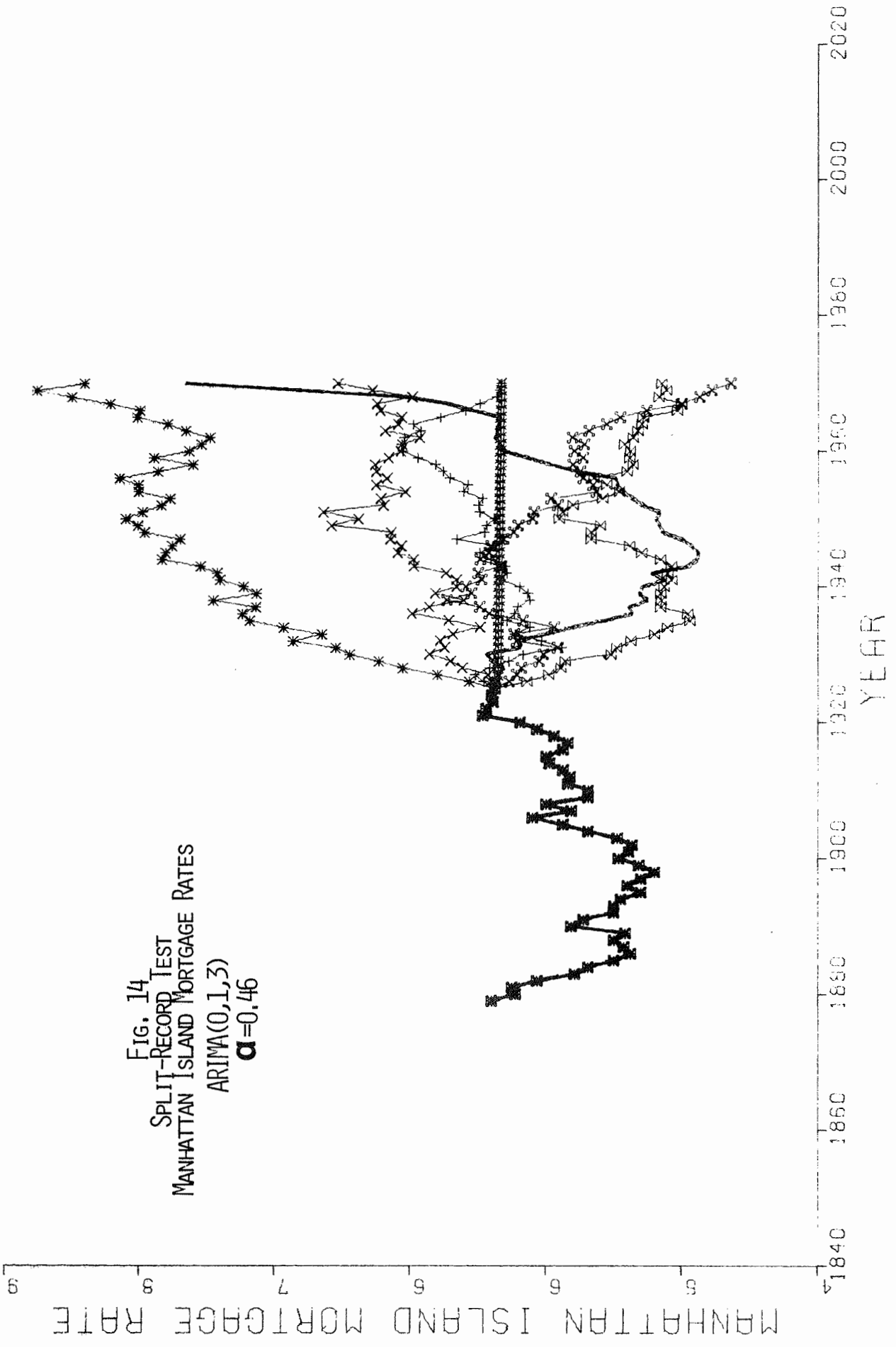


FIG. 12
 SPLIT-RECORD TEST
 WHOLESale PRICE INDEX
 ARIMA(0,1,1)
 $\alpha = 0.92$

WHOLESale PRICE INDEX (1967=100)

YEAR





MANHATTAN ISLAND MORTGAGE RATE

1840 1860 1880 1900 1920 1940 1960 1980 2000 2020

YEAR

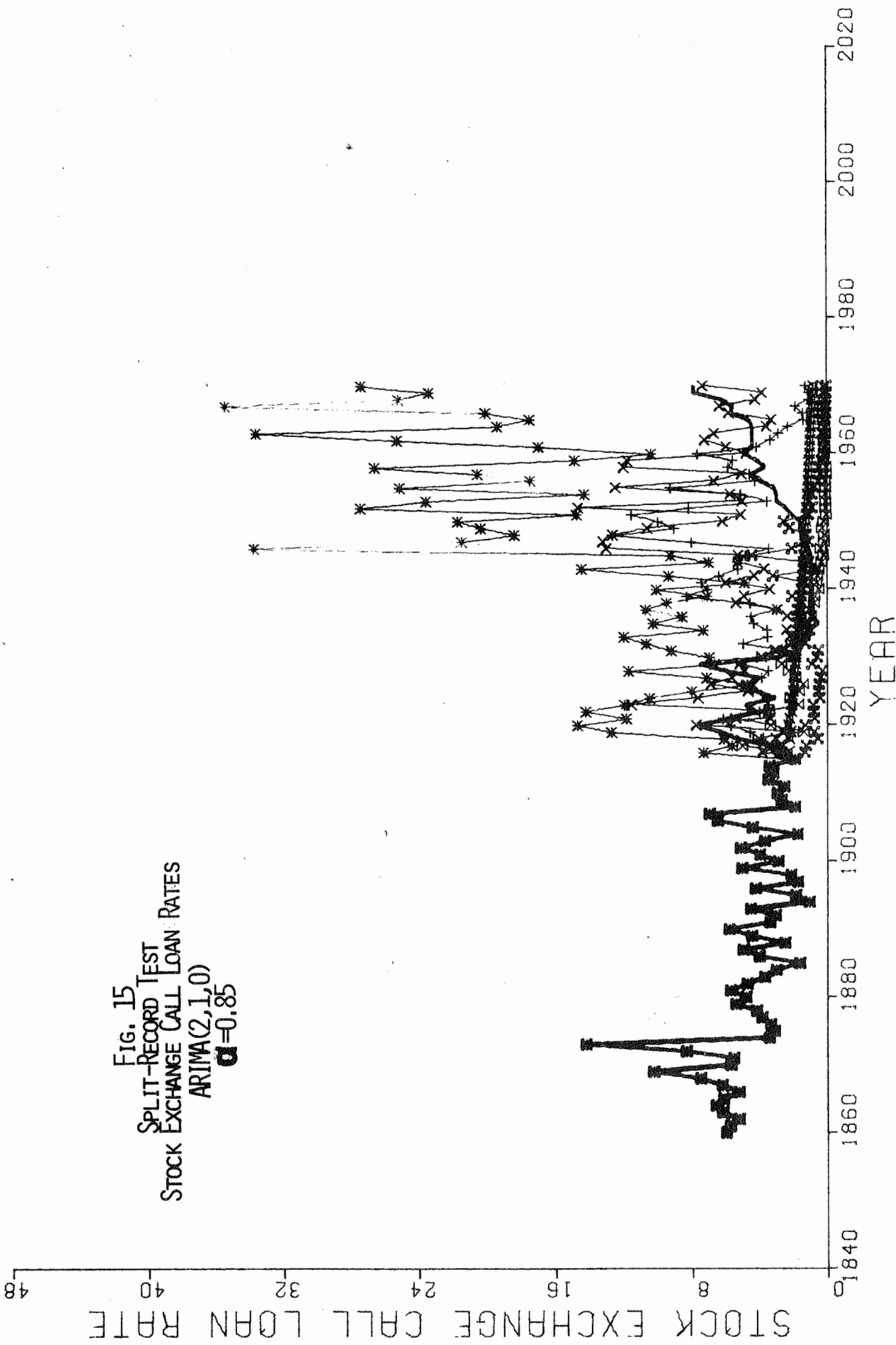


Fig. 15
 SPLIT-RECORD TEST
 STOCK EXCHANGE CALL LOAN RATES
 ARIMA(2,1,0)
 $\alpha = 0.85$

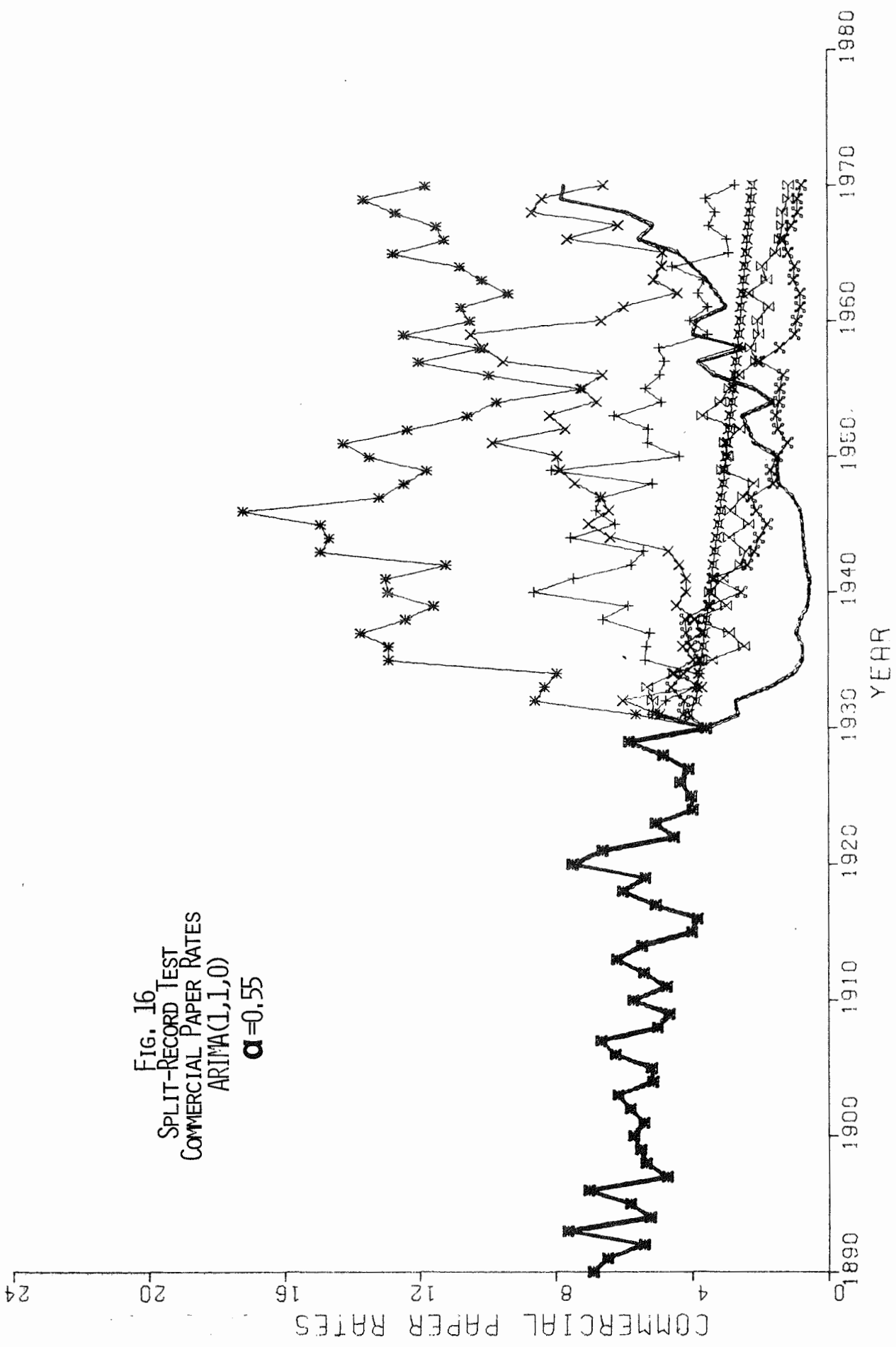


FIG. 16
 SPLIT-RECORD TEST
 COMMERCIAL PAPER RATES
 ARIMA(1,1,0)
 $\alpha = 0.55$

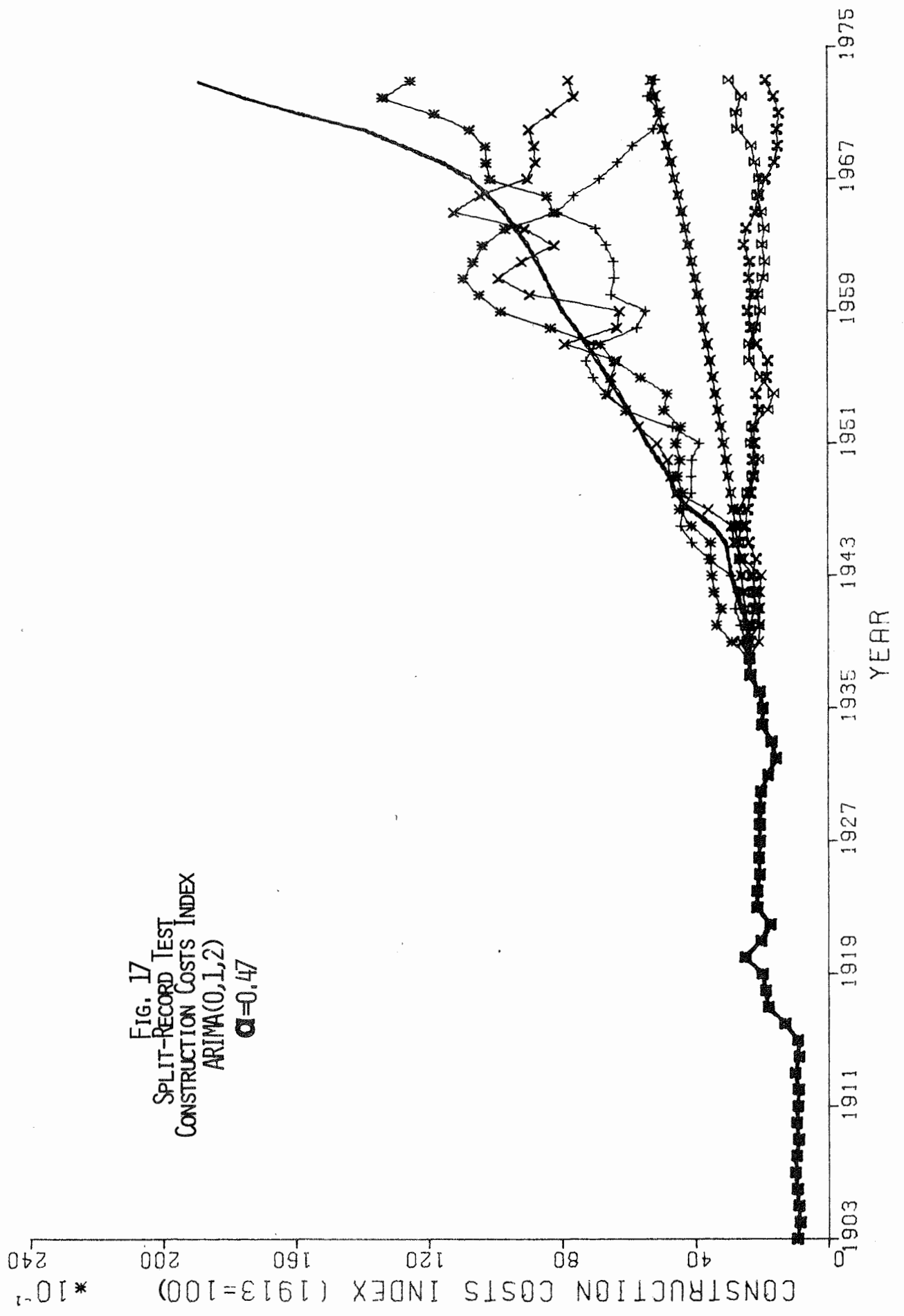
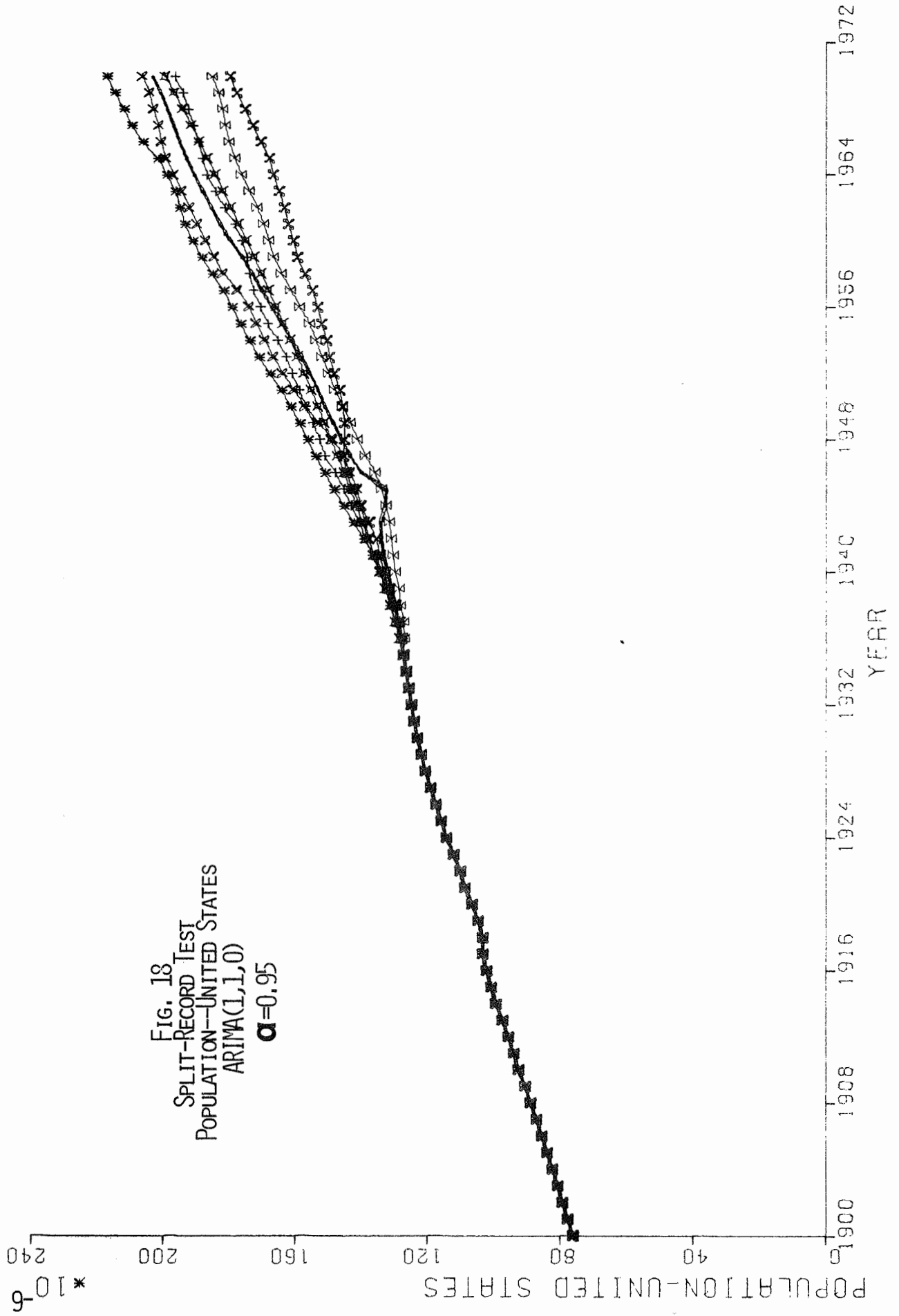
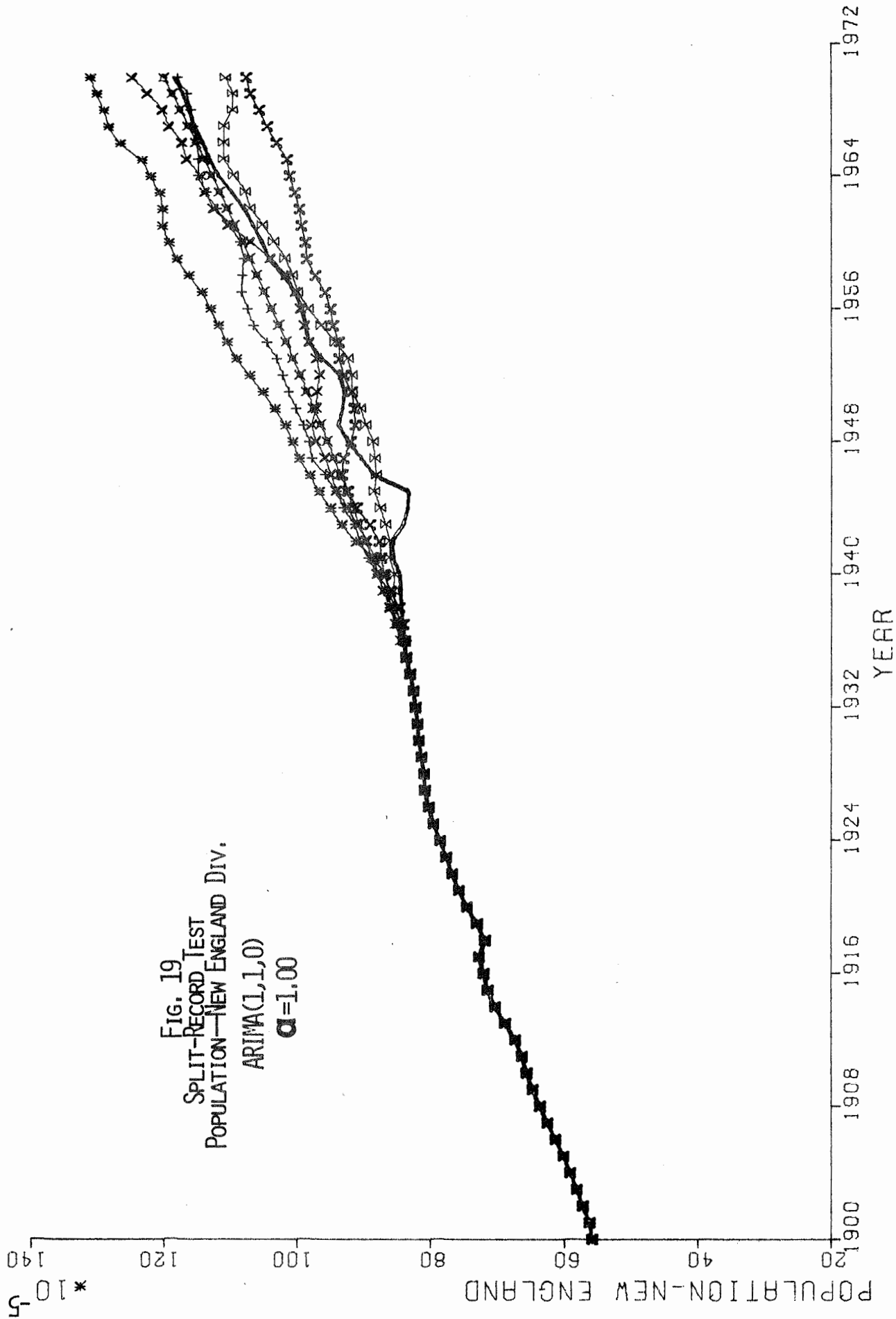


FIG. 17
 SPLIT-RECORD TEST
 CONSTRUCTION COSTS INDEX
 ARIMA(0,1,2)
 $\alpha = 0.47$





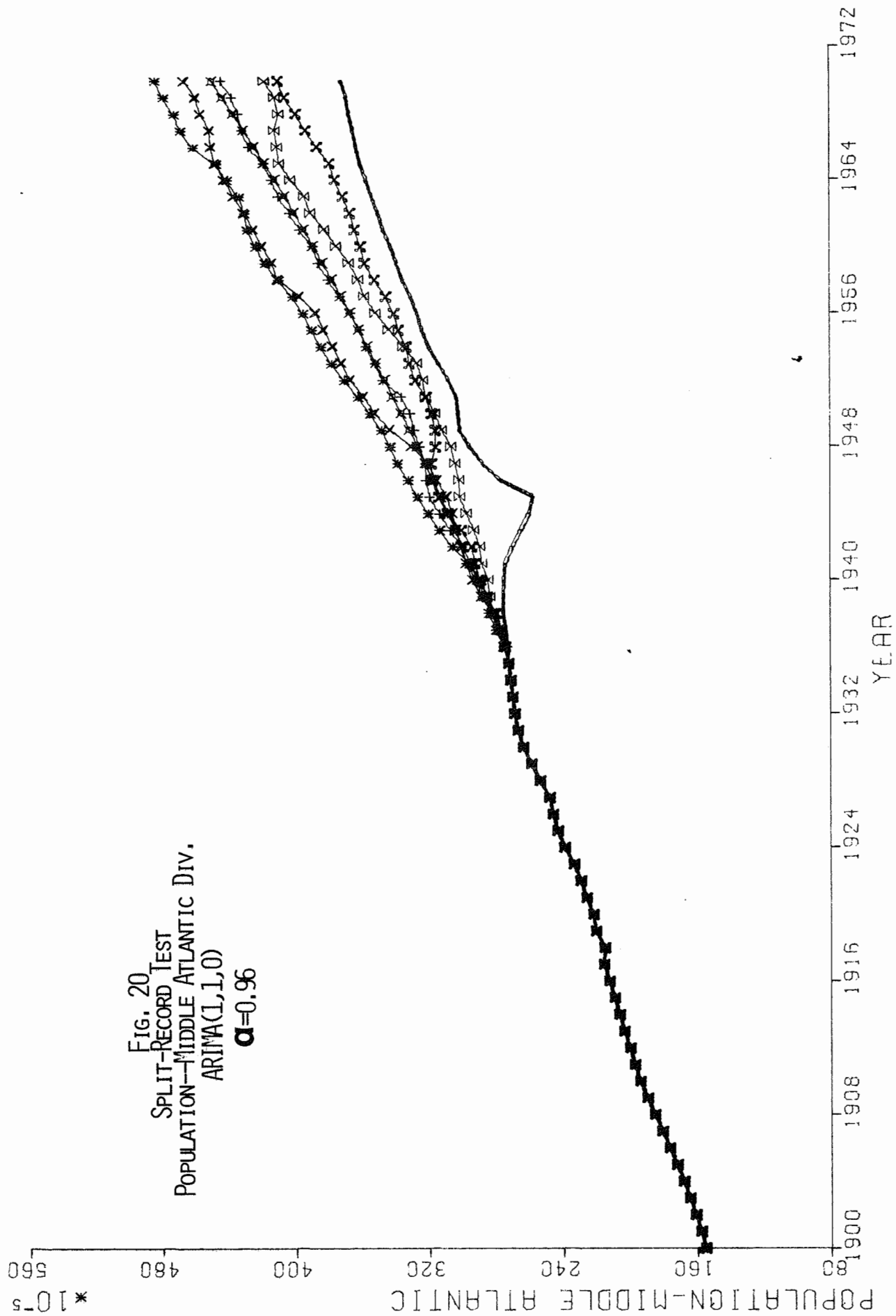
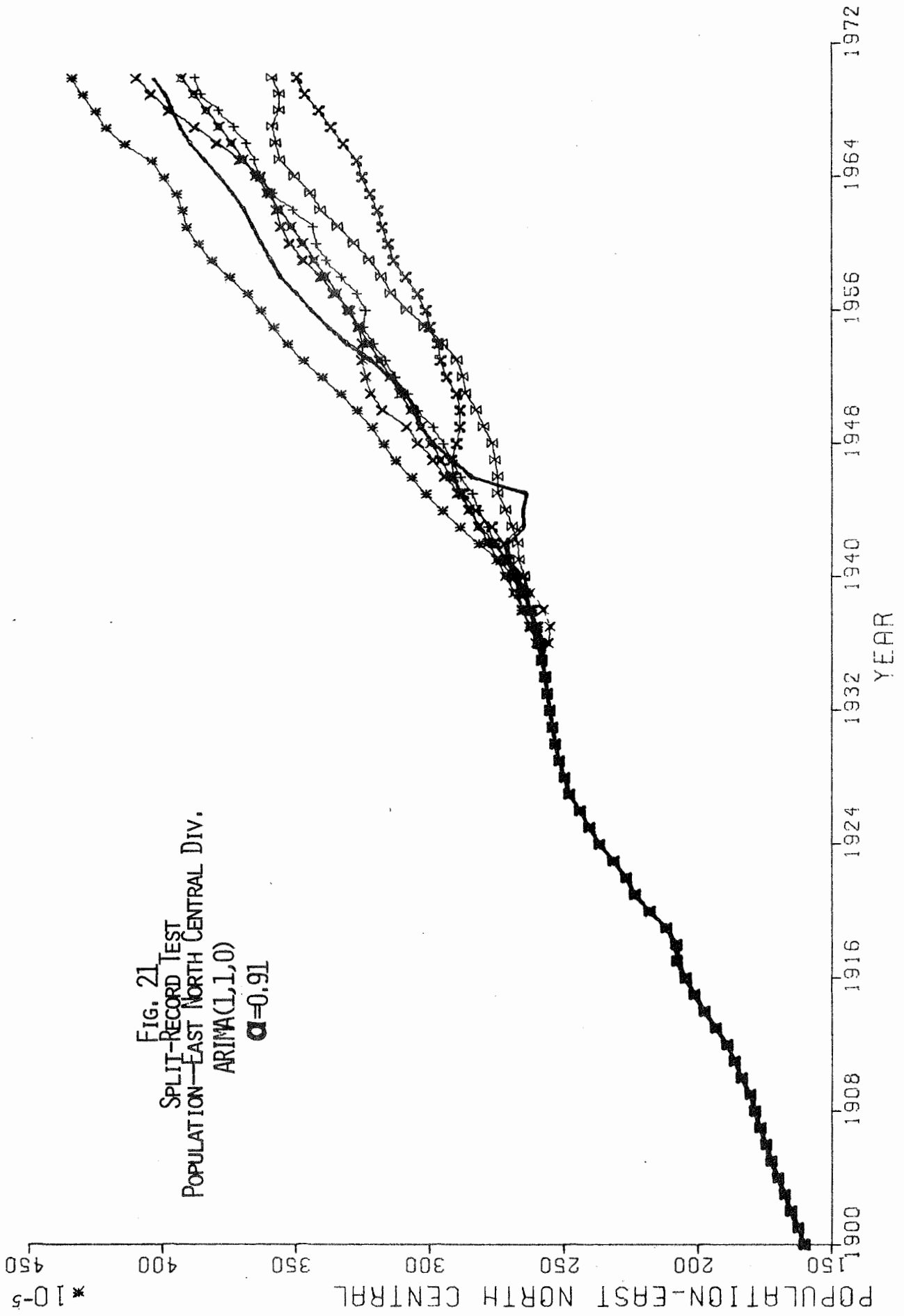


FIG. 20
 SPLIT-RECORD TEST
 POPULATION-MIDDLE ATLANTIC DIV.
 ARIMA(1,1,0)
 $\alpha=0.96$



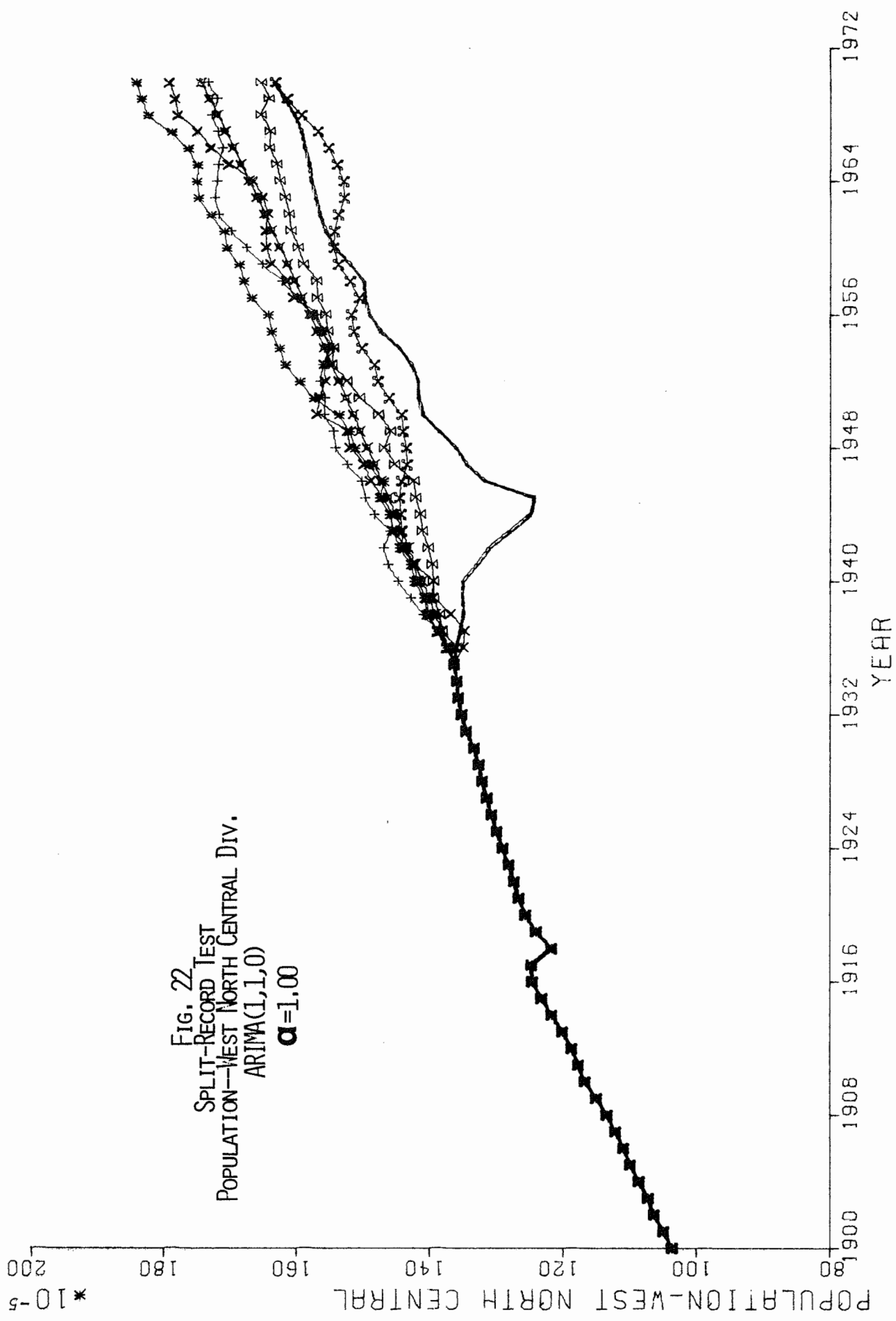


Fig. 22
 SPLIT-RECORD TEST
 POPULATION--WEST NORTH CENTRAL DIV.
 ARIMA(1,1,0)
 $\alpha=1.00$

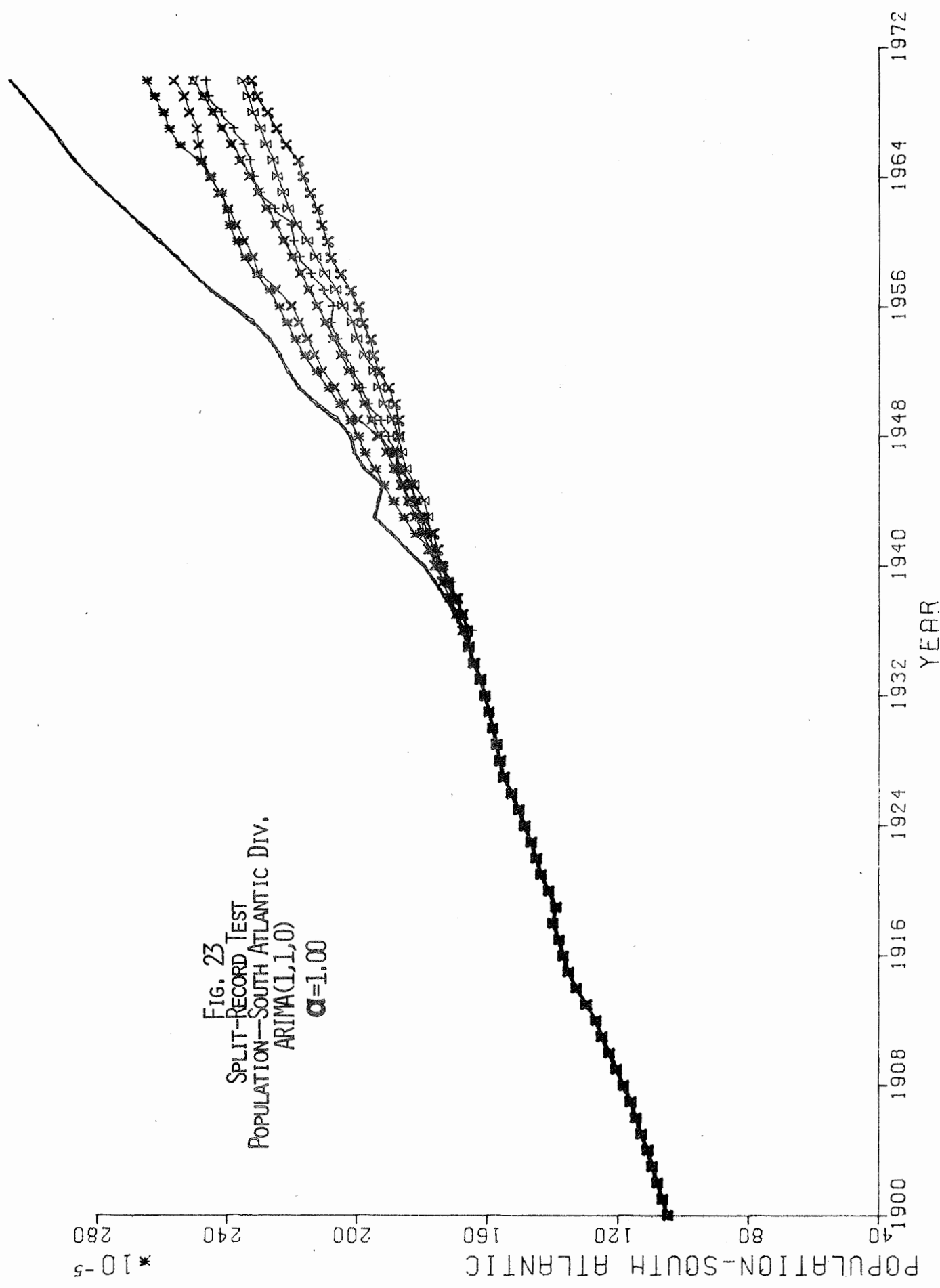


FIG. 23
 SPLIT-RECORD TEST
 POPULATION-SOUTH ATLANTIC DIV.
 ARIMA(1,1,0)
 $\alpha=1.00$

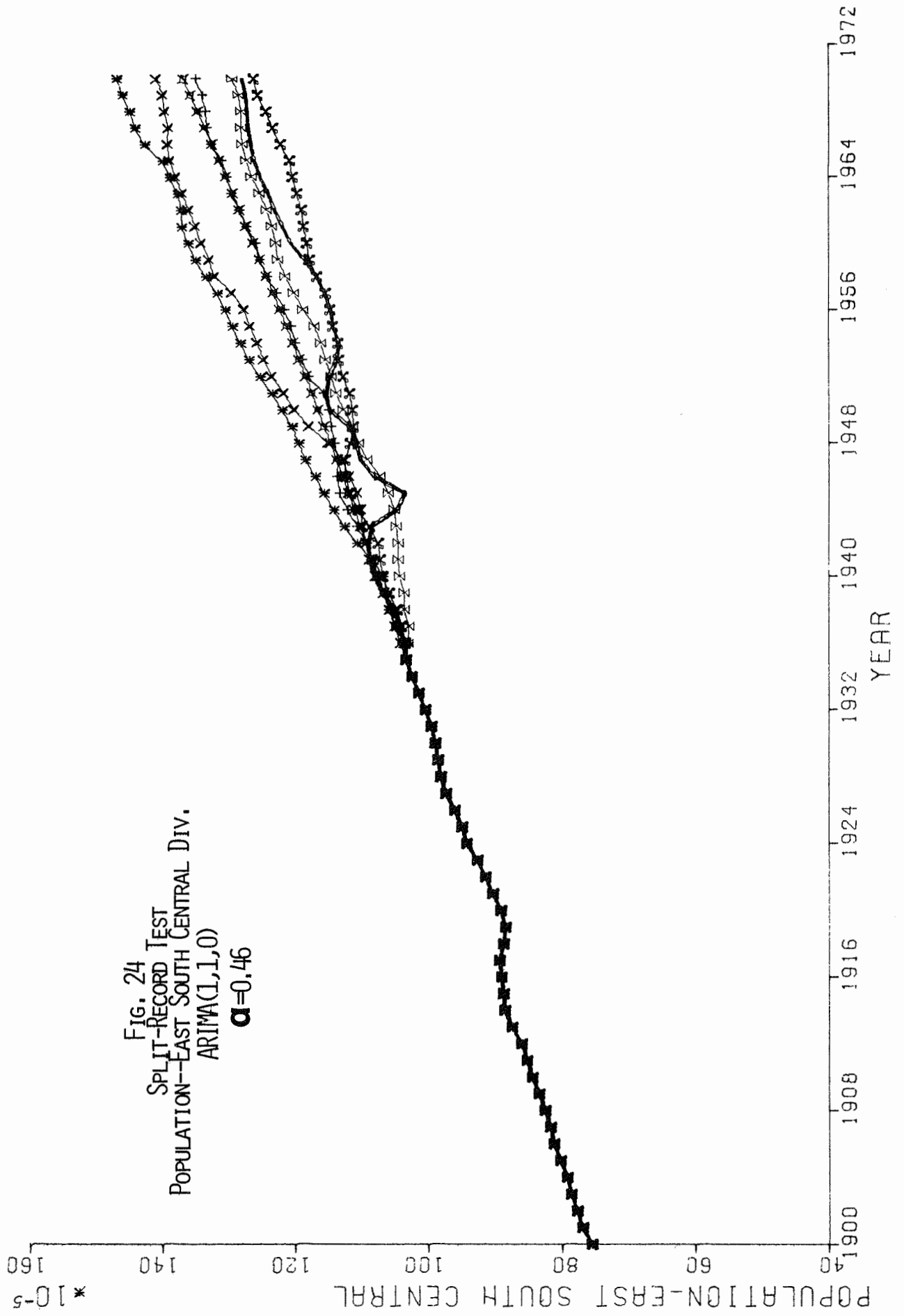


FIG. 24
 SPLIT-RECORD TEST
 POPULATION--EAST SOUTH CENTRAL DIV.
 ARIMA(1,1,0)
 $\alpha = 0.46$

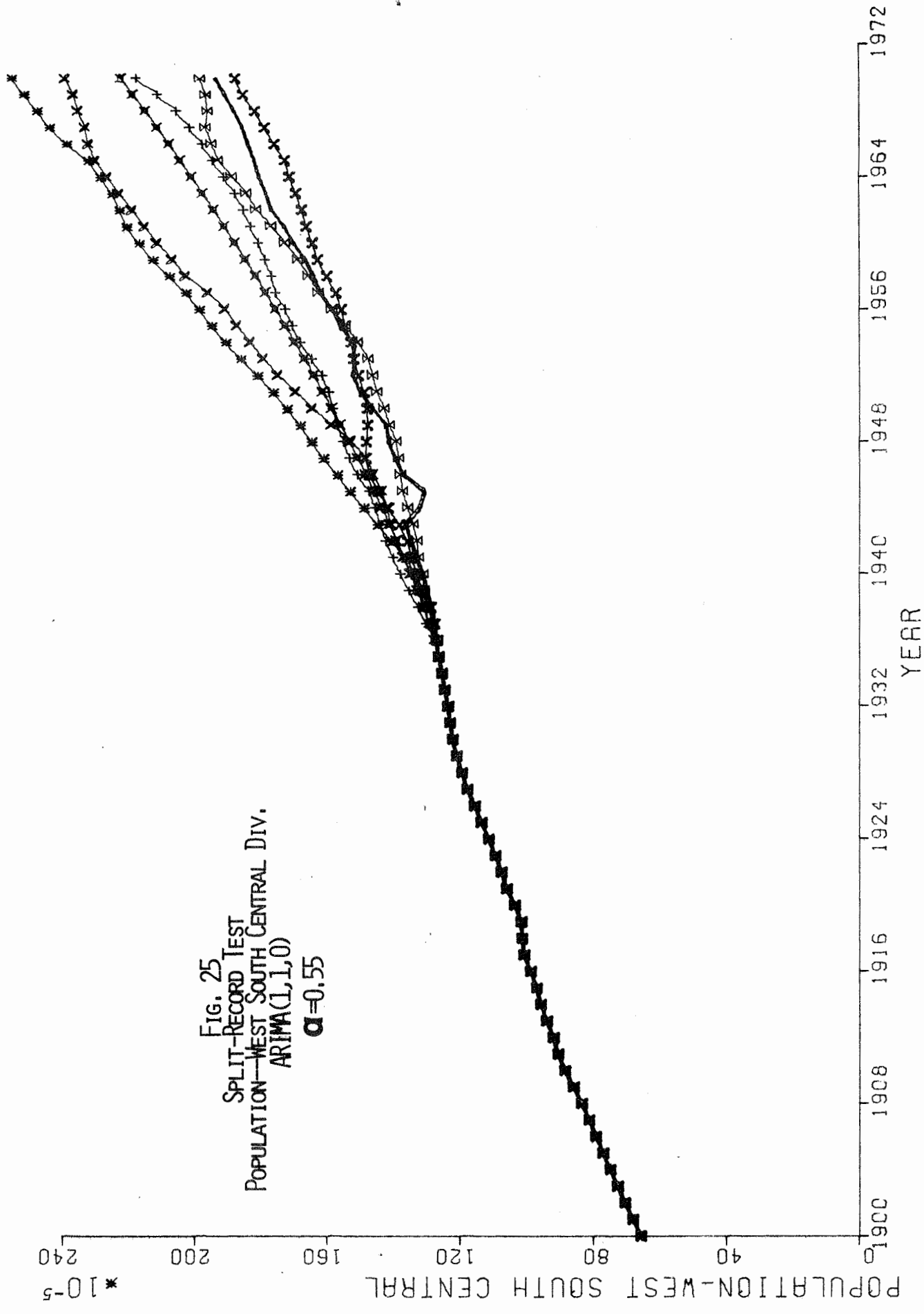


FIG. 25
 SPLIT-RECORD TEST
 POPULATION—WEST SOUTH CENTRAL DIV.
 ARIMA(1,1,0)
 $\alpha = 0.55$

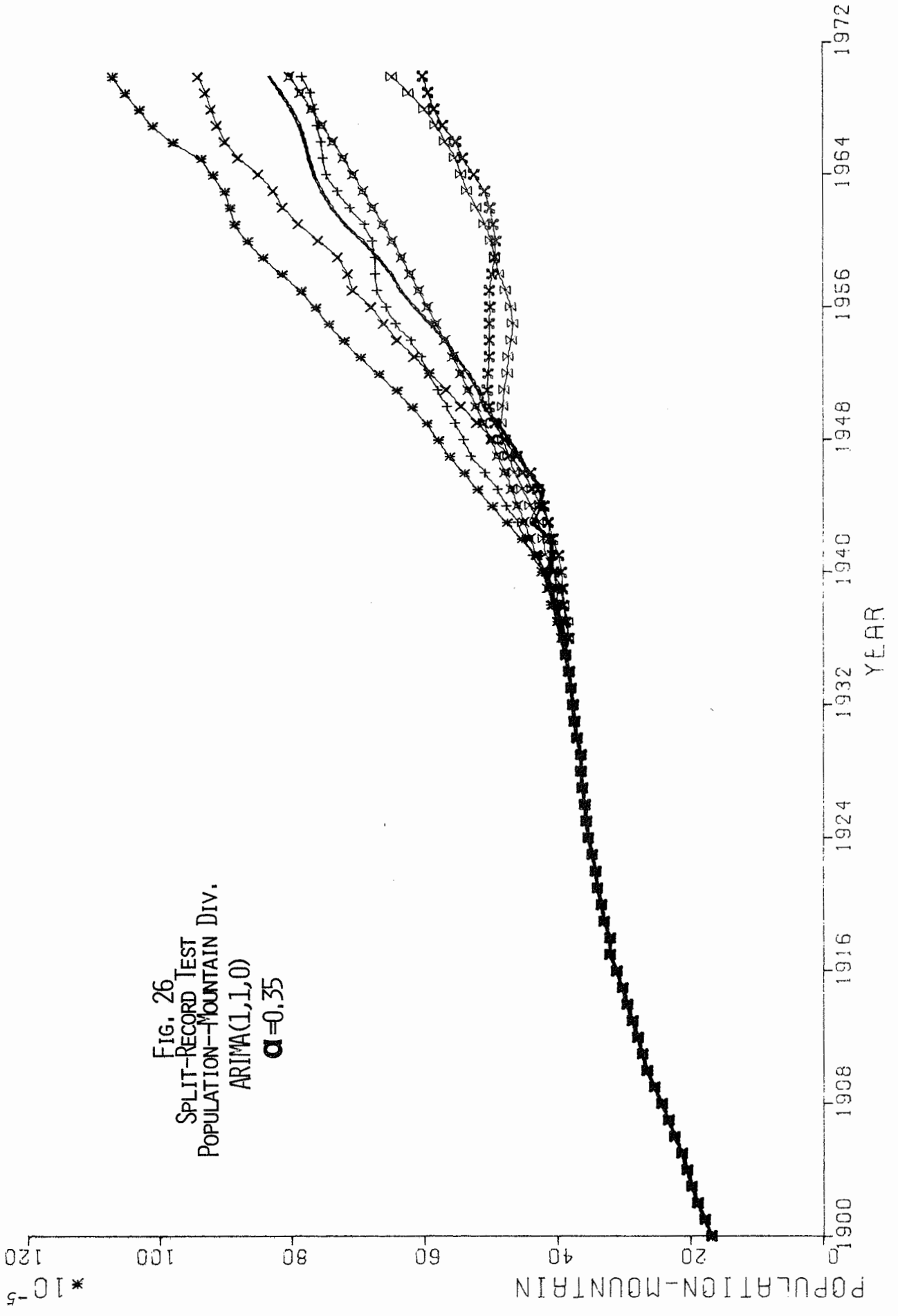


FIG. 26
 SPLIT-RECORD TEST
 POPULATION-MOUNTAIN DIV.
 ARIMA(1,1,0)
 $\alpha = 0.35$

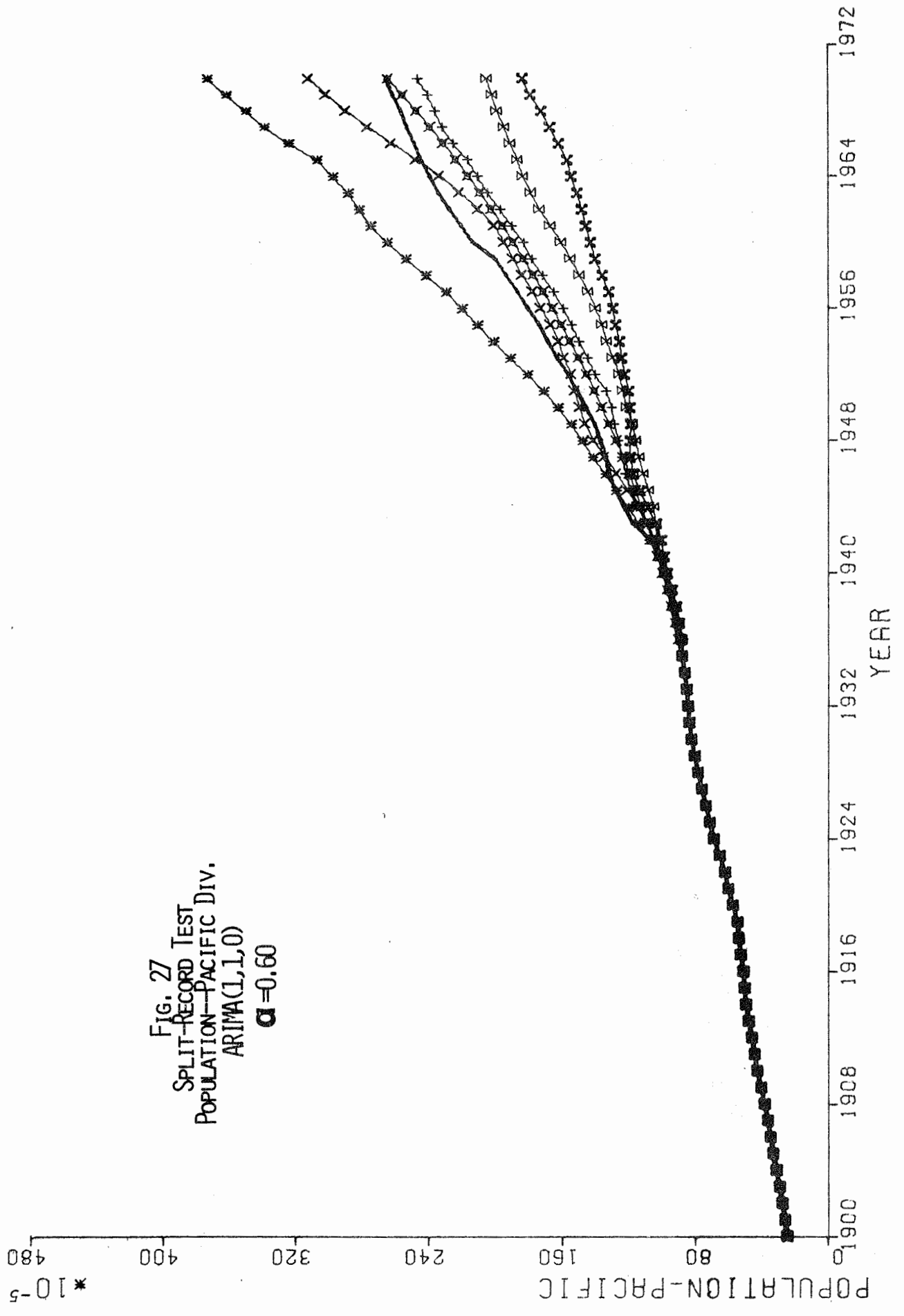


FIG. 27
 SPLIT-RECORD TEST
 POPULATION-PACIFIC DIV.
 ARIMA(1,1,0)
 $\alpha = 0.60$

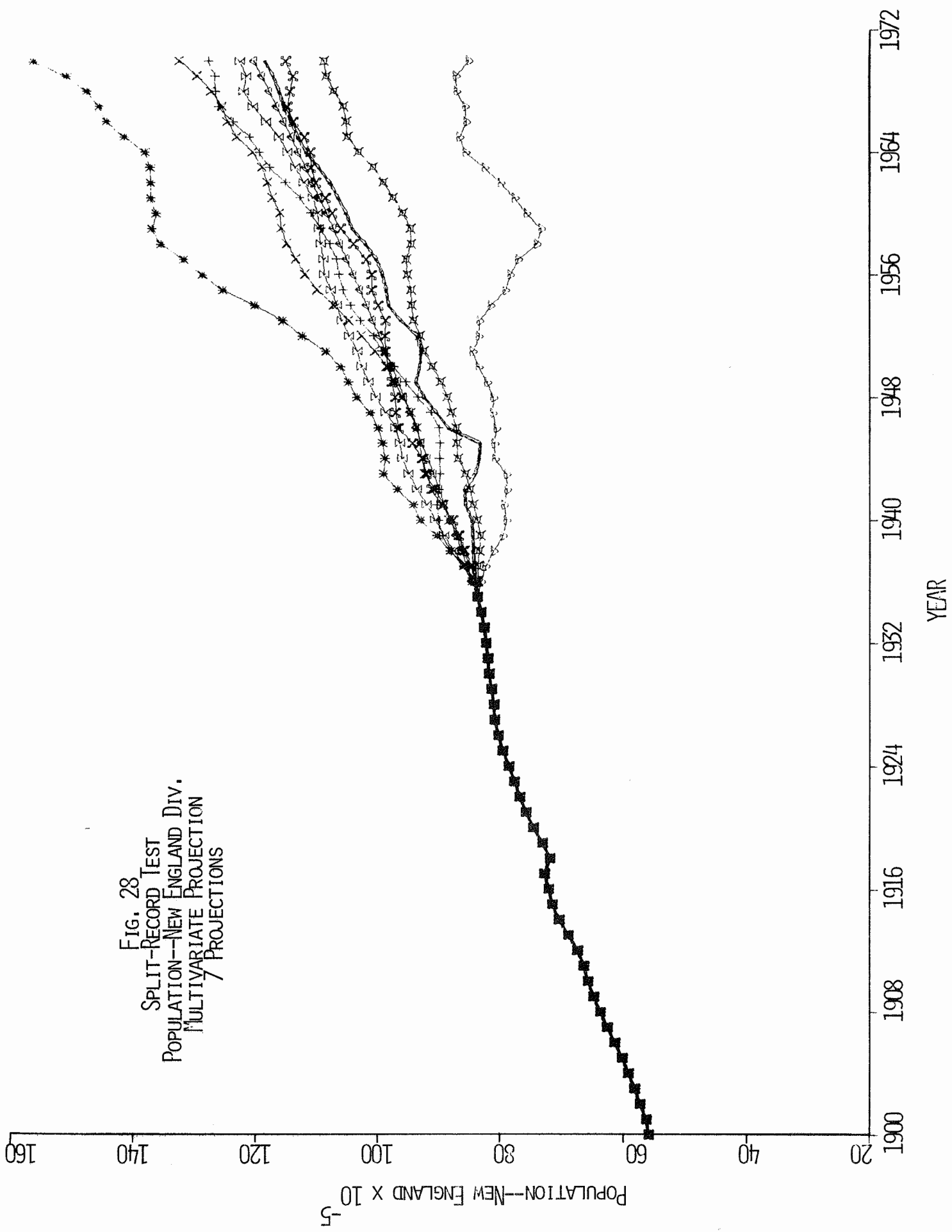


Fig. 28.
 SPLIT-RECORD TEST
 POPULATION--NEW ENGLAND DIV.
 MULTIVARIATE PROJECTIONS

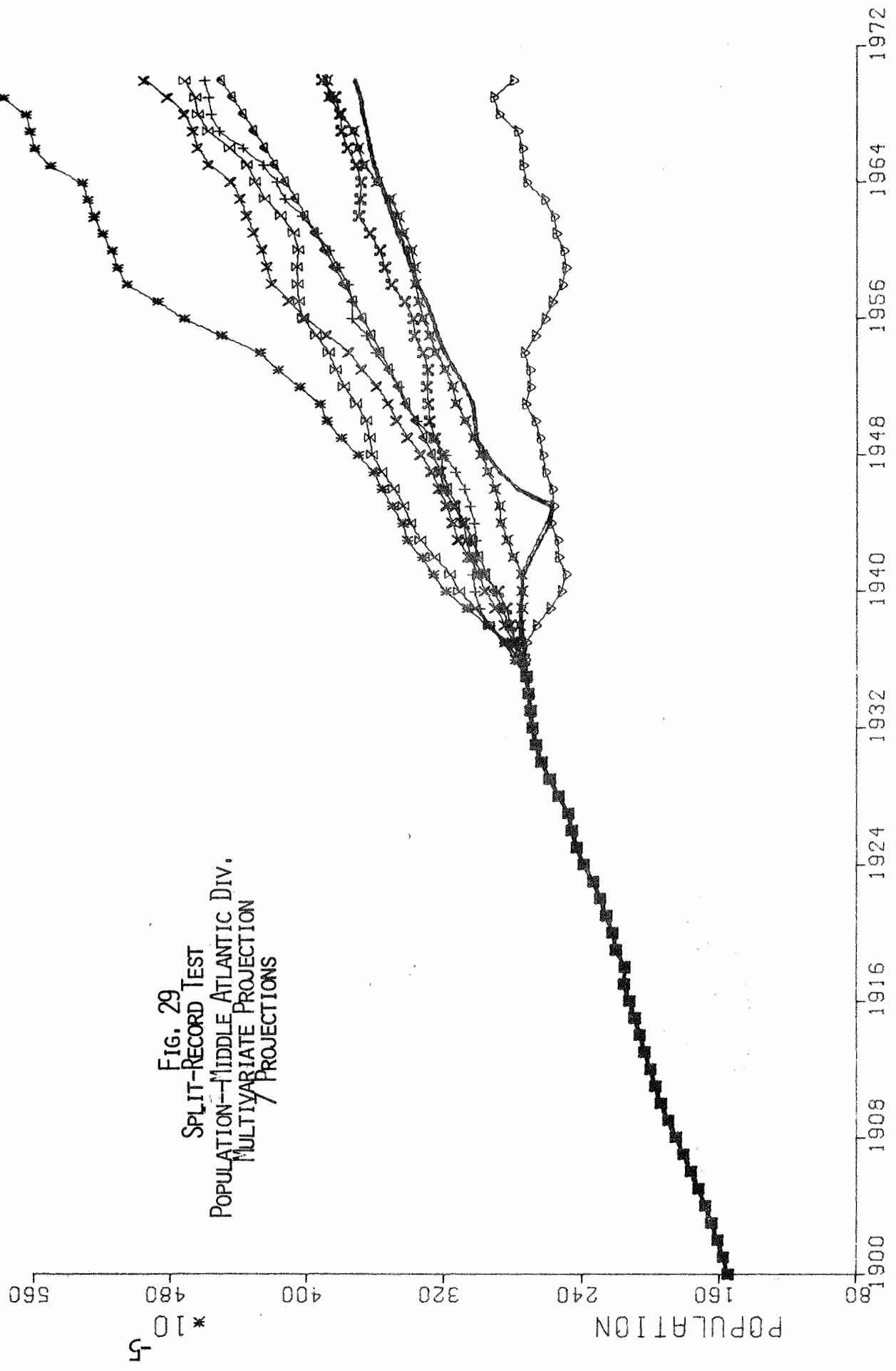


FIG. 29
 SPLIT-RECORD TEST
 POPULATION--MIDDLE ATLANTIC DIV.
 MULTIVARIATE PROJECTION
 PROJECTIONS

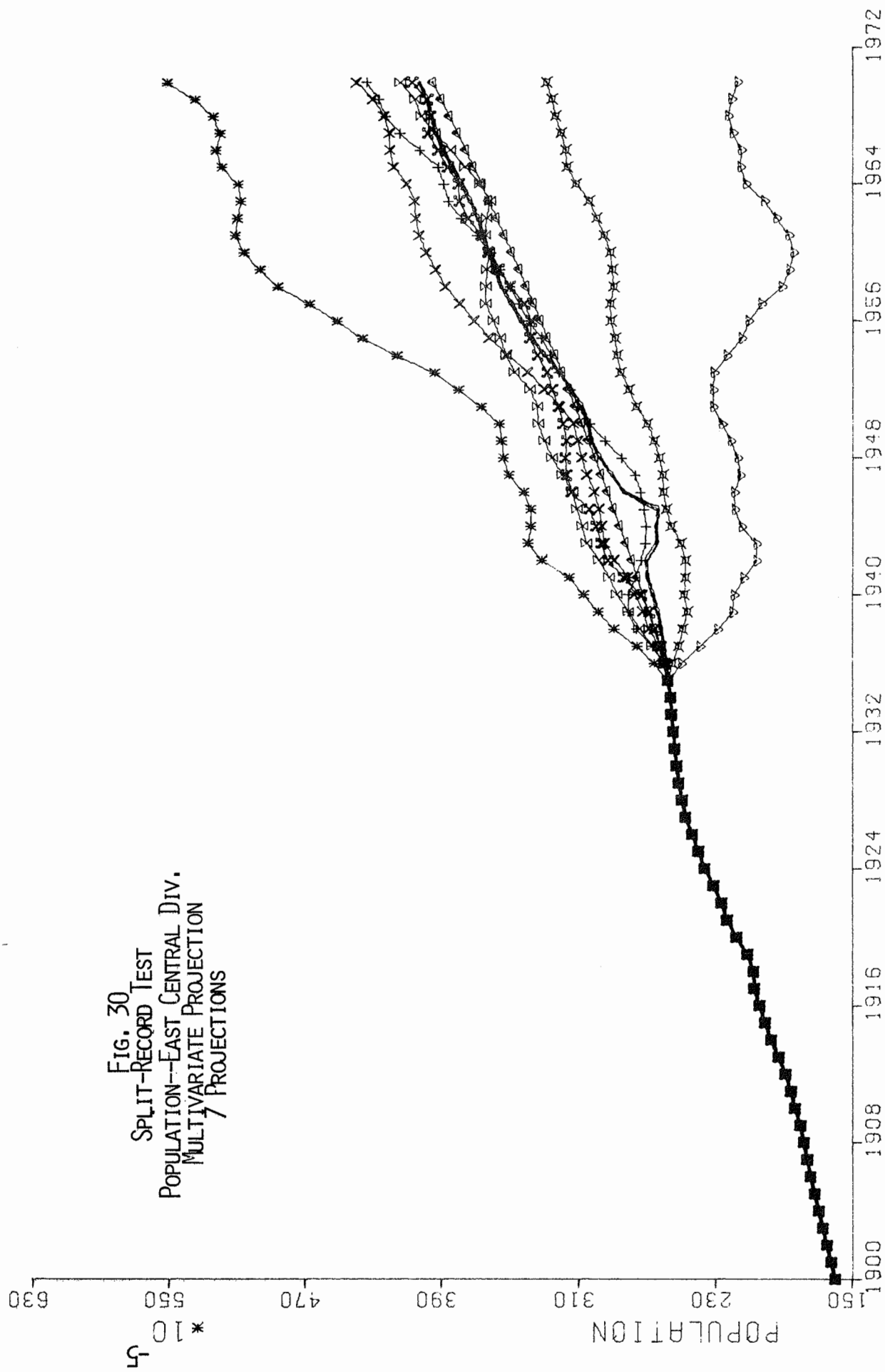


FIG. 30
 SPLIT-RECORD TEST
 POPULATION--EAST CENTRAL DIV.
 MULTIVARIATE PROJECTION
 / PROJECTIONS

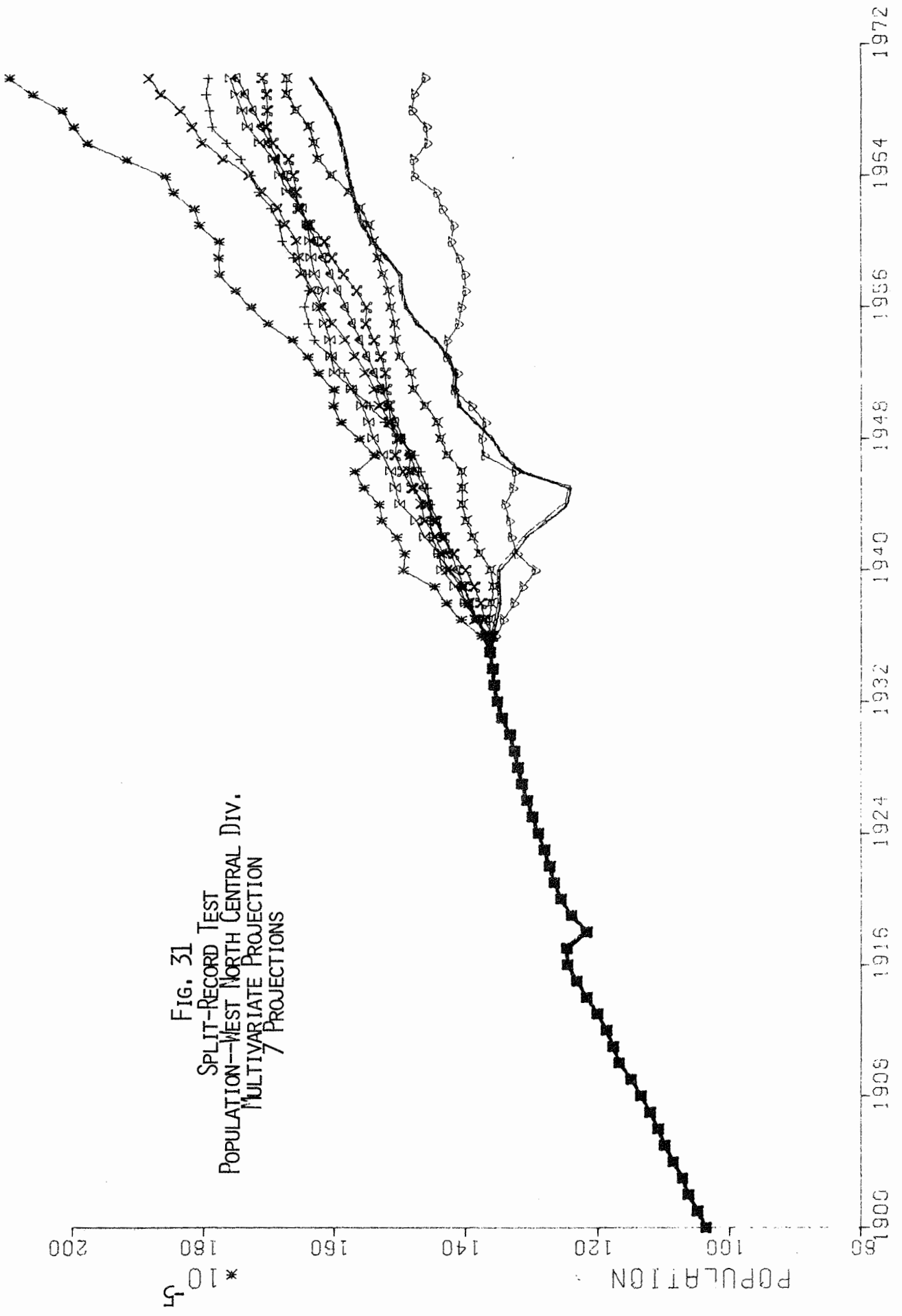


FIG. 31
 SPLIT-RECORD TEST
 POPULATION--WEST NORTH CENTRAL DIV.
 MULTIVARIATE PROJECTION
 7 PROJECTIONS

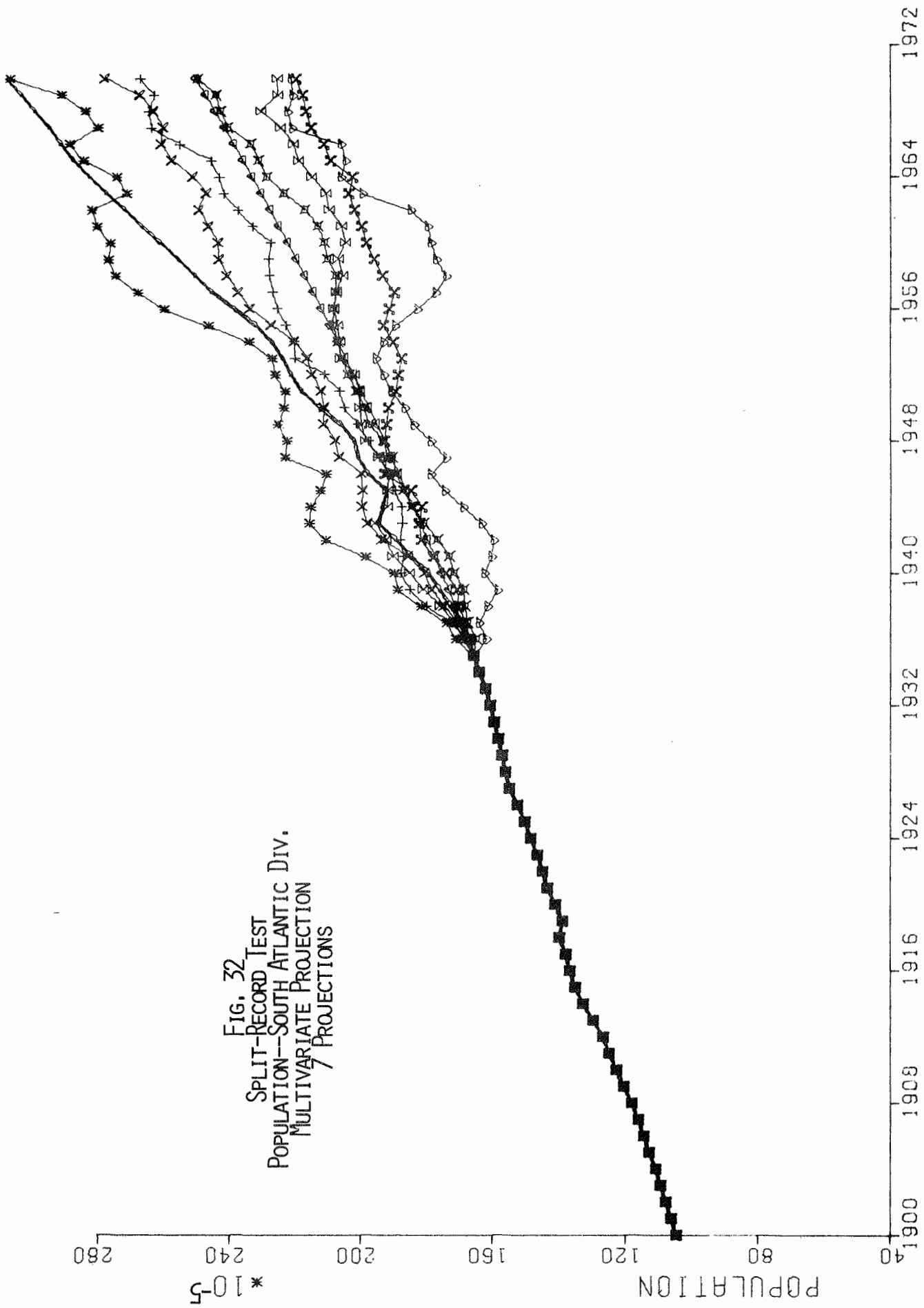


FIG. 32
 SPLIT-RECORD TEST
 POPULATION--SOUTH ATLANTIC DIV.
 MULTIVARIATE PROJECTION
 / PROJECTIONS

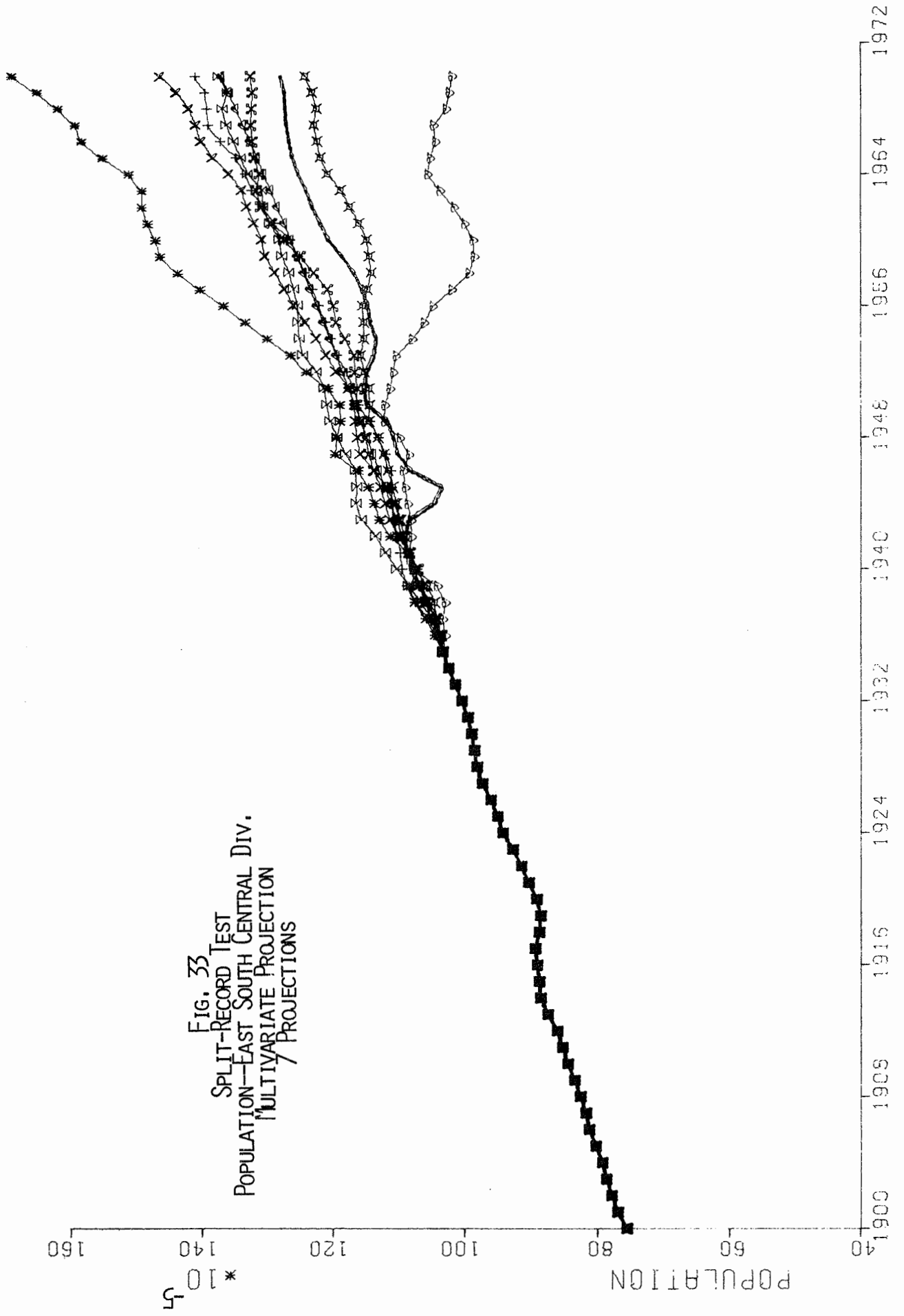
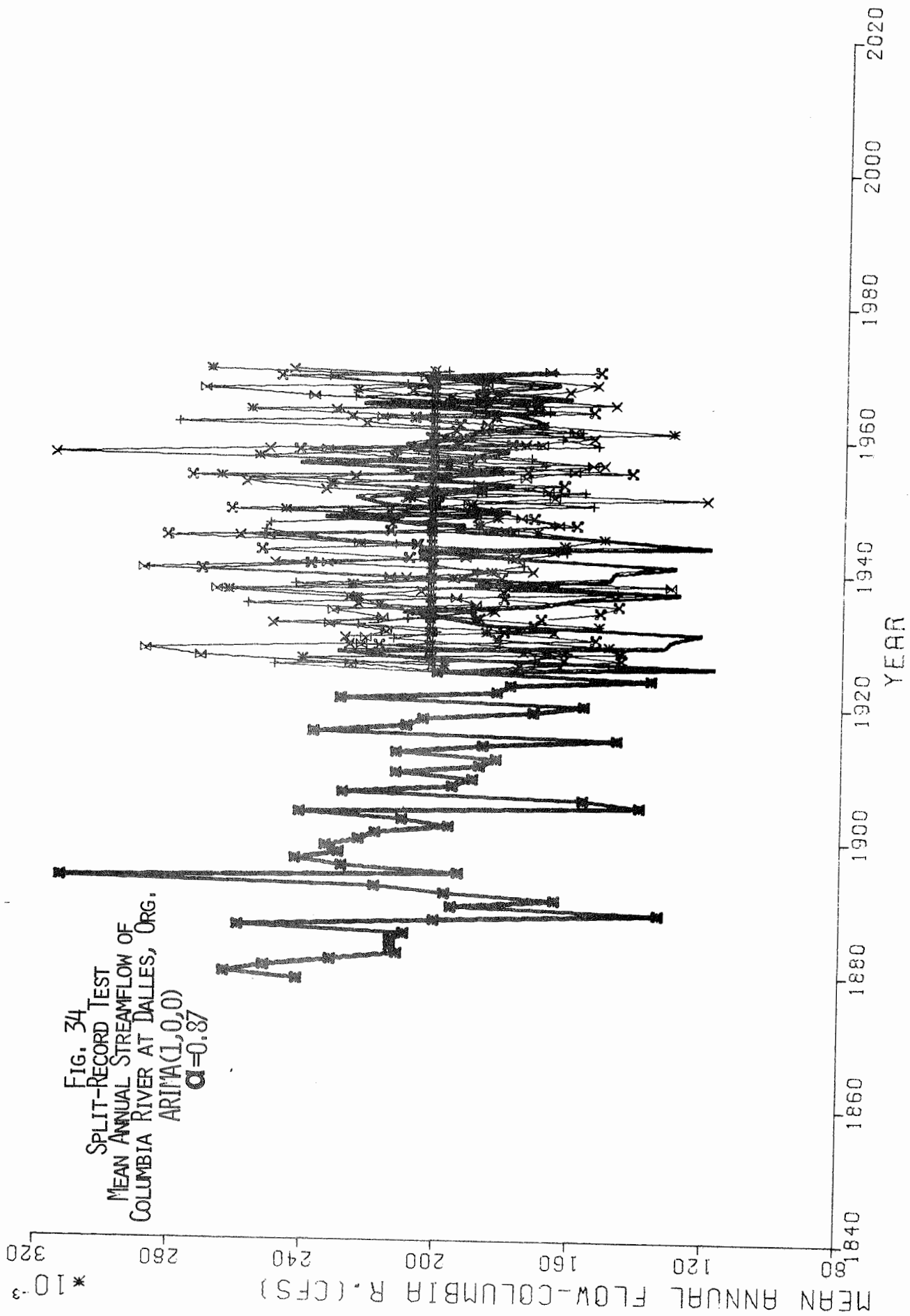
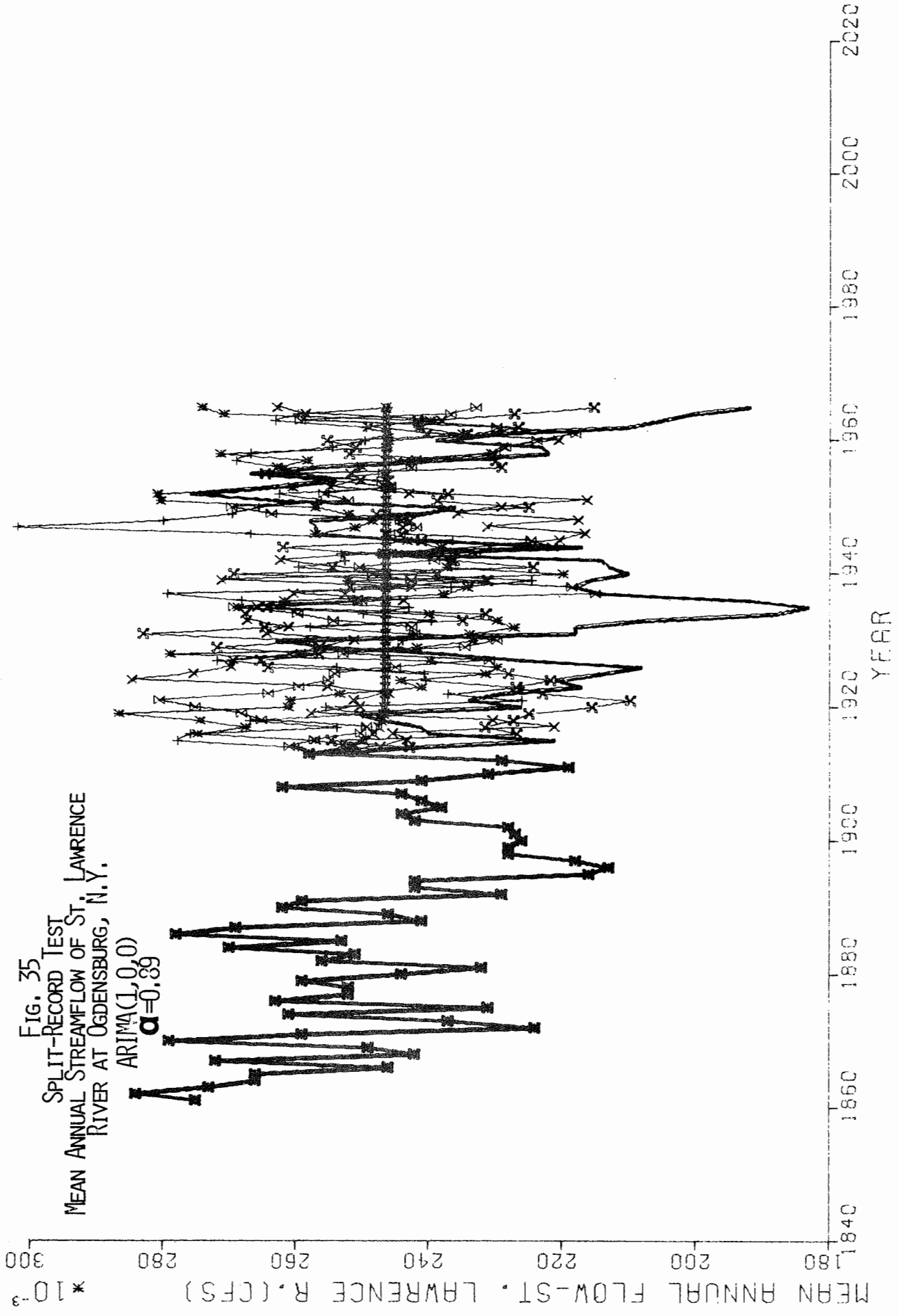


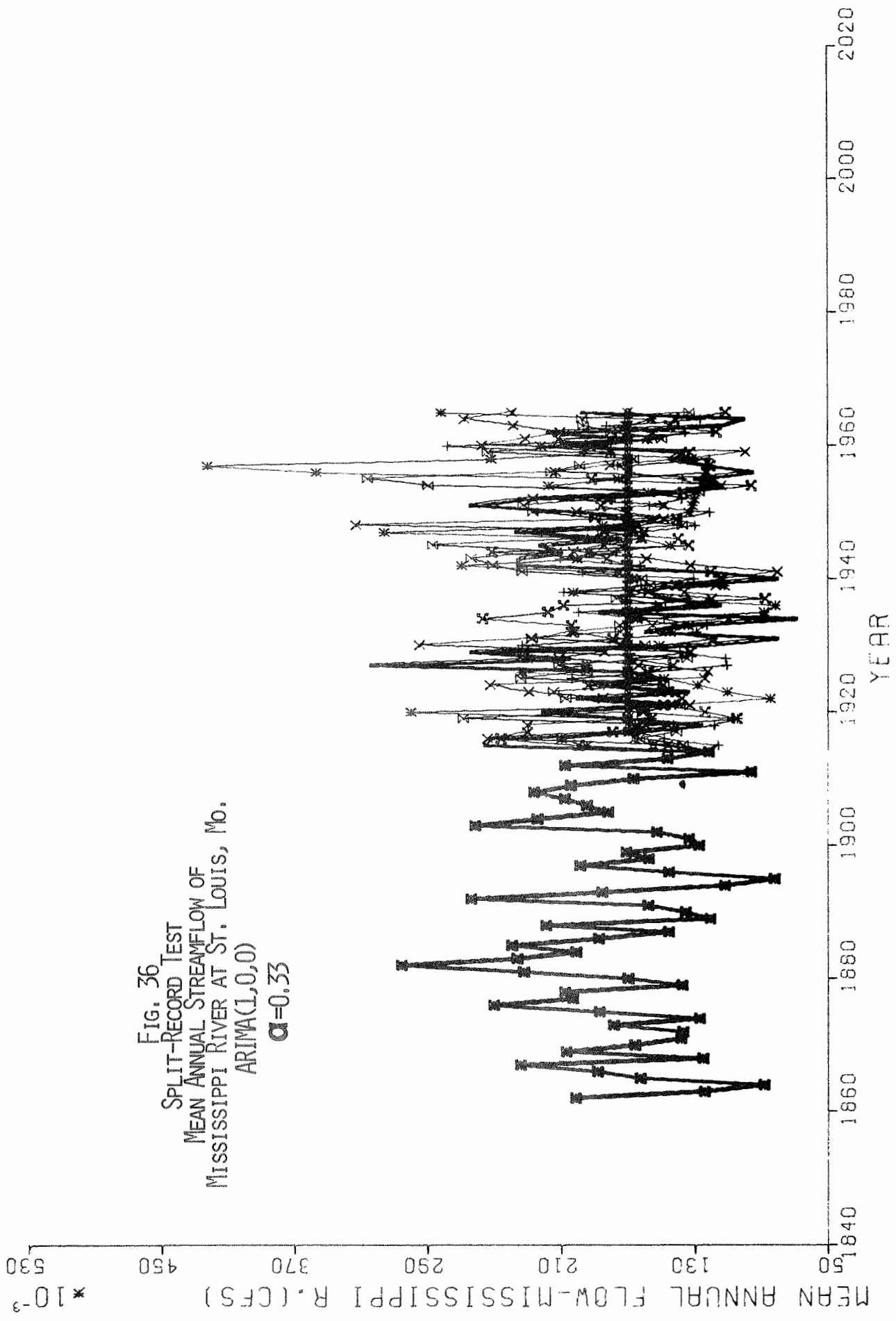
FIG. 33
 SPLIT-RECORD TEST
 POPULATION—EAST SOUTH CENTRAL DIV.
 MULTIVARIATE PROJECTIONS





MEAN ANNUAL FLOW-ST. LAWRENCE R. (CFS) * 10⁻³

1840 1860 1880 1900 1920 1940 1960 1980 2000 2020
YEAR



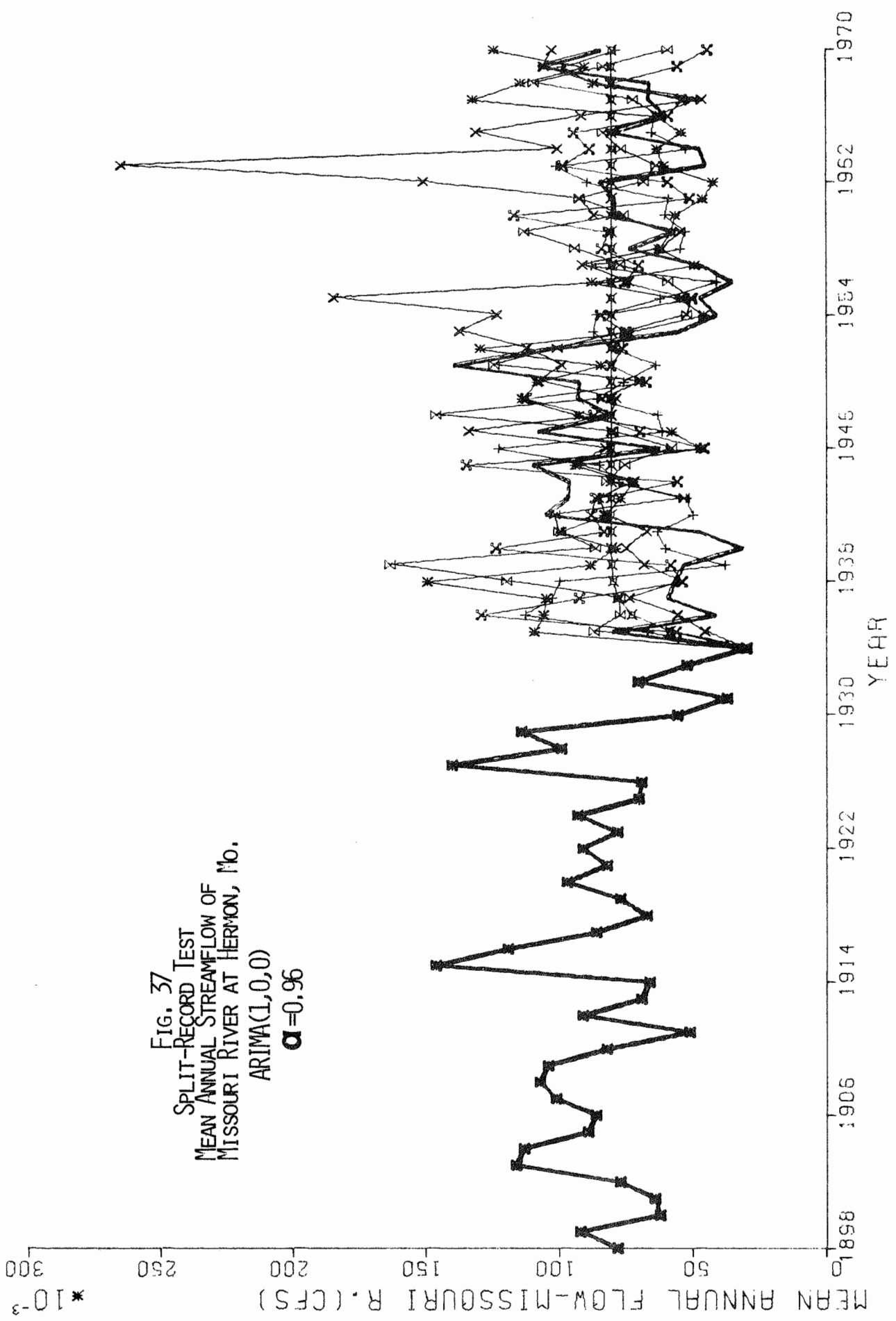
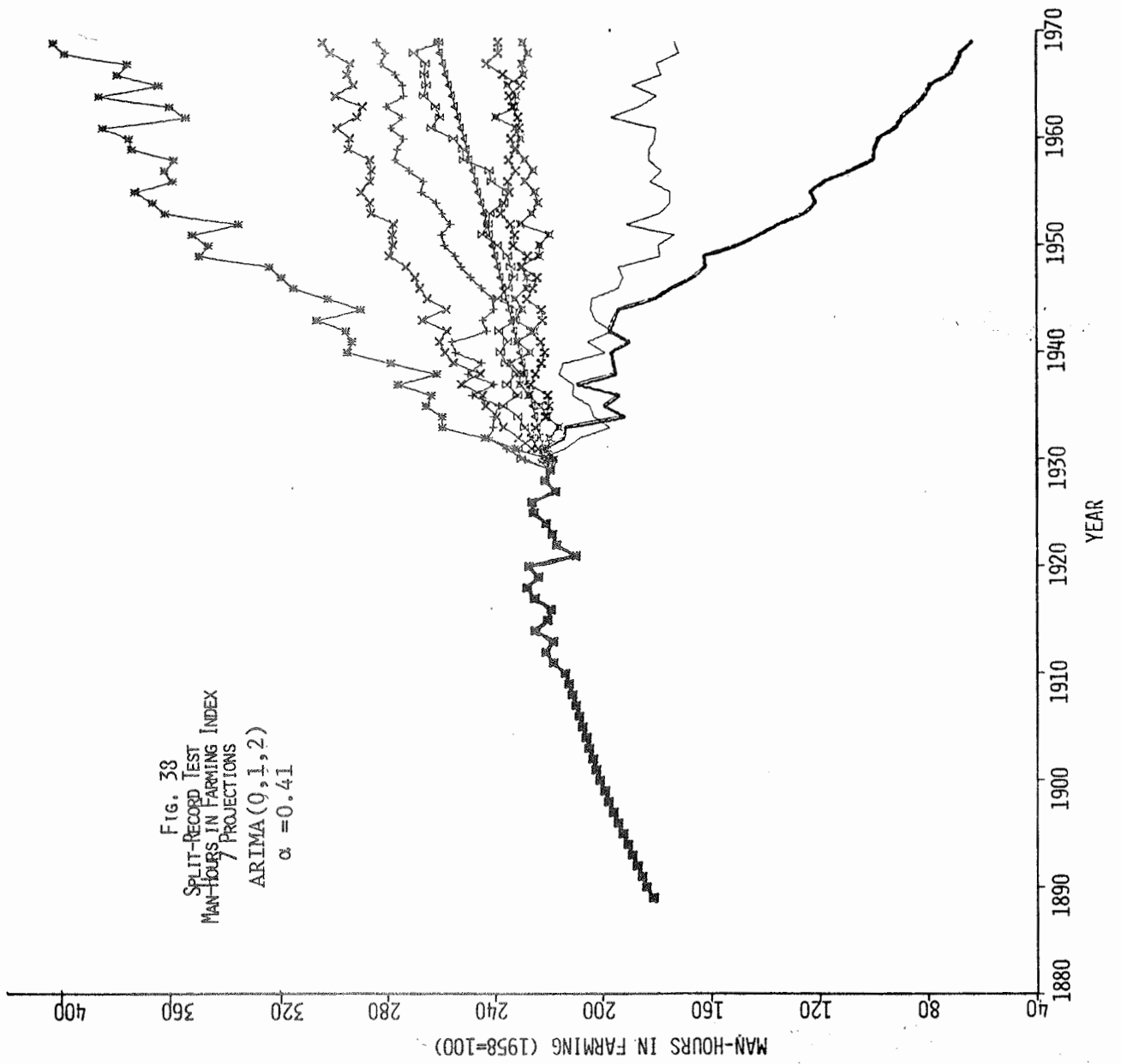


FIG. 37
 SPLIT-RECORD TEST
 MEAN ANNUAL STREAMFLOW OF
 MISSOURI RIVER AT HERMON, Mo.
 ARIMA(1,0,0)
 $\alpha = 0.96$



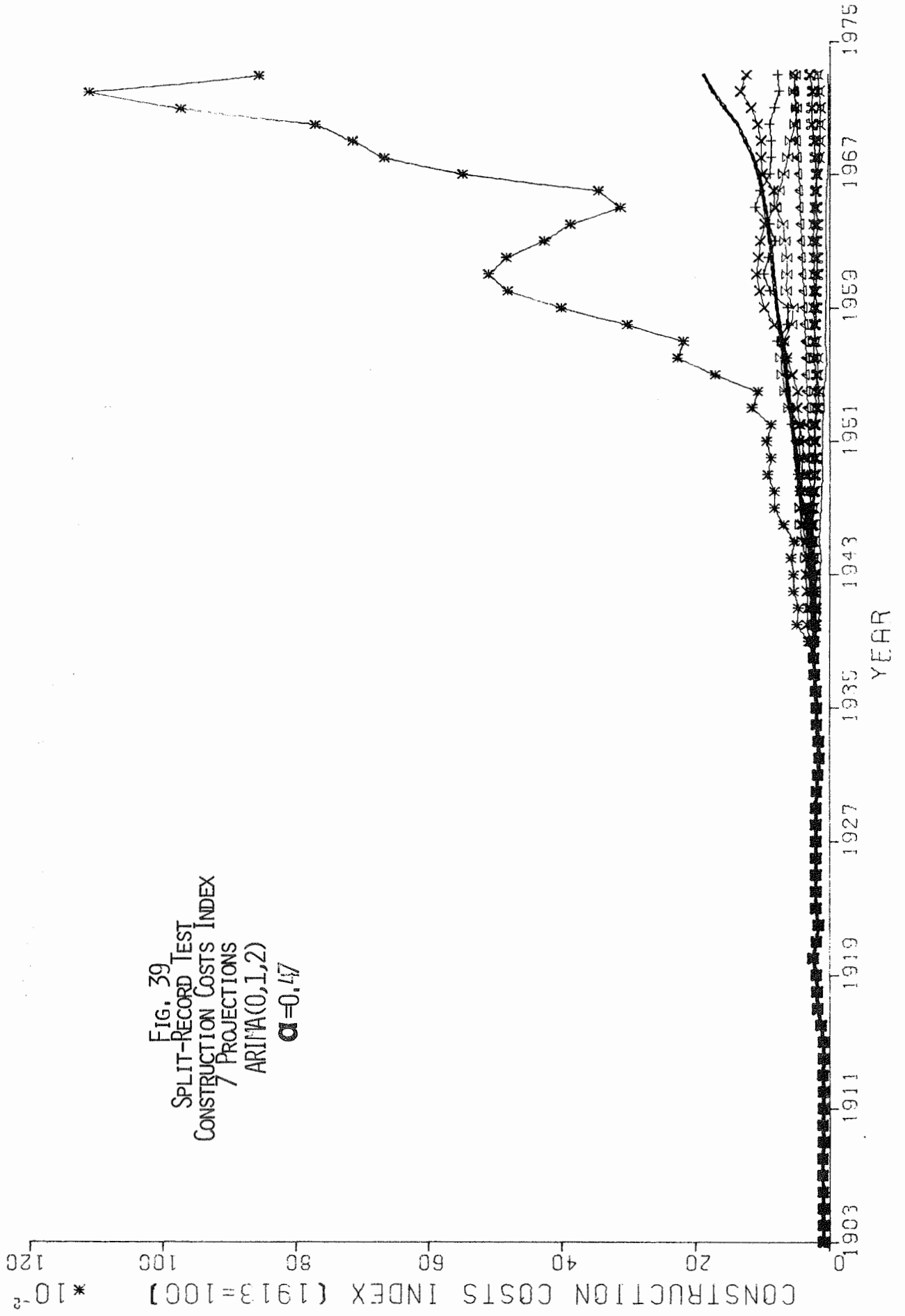


Fig. 39
 SPLIT-RECORD TEST
 CONSTRUCTION COSTS INDEX
 / PROJECTIONS
 ARIMA(0,1,2)
 $\alpha = 0.47$

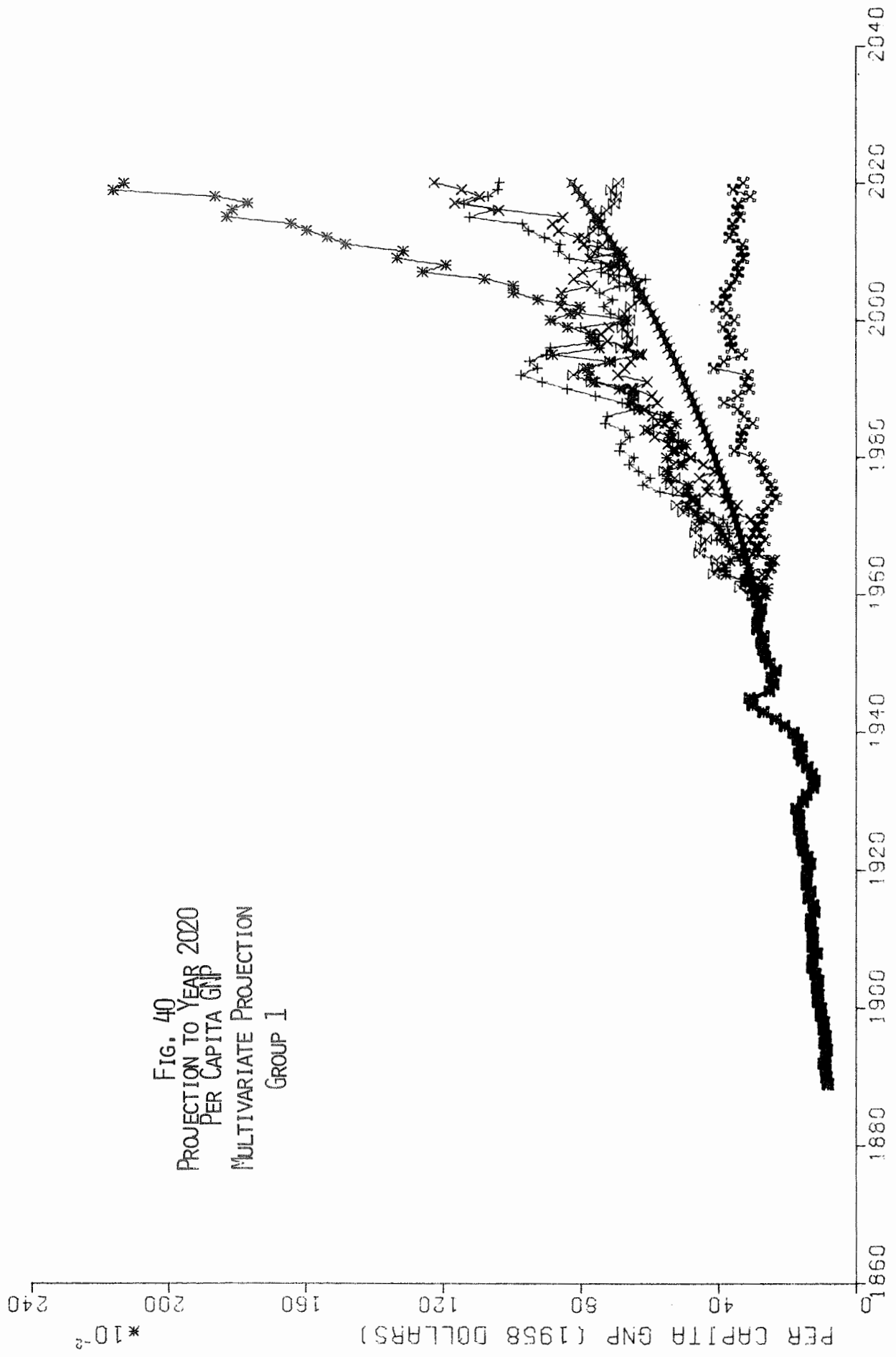


FIG. 40
 PROJECTION TO YEAR 2020
 PER CAPITA GNP
 MULTIVARIATE PROJECTION
 GROUP 1

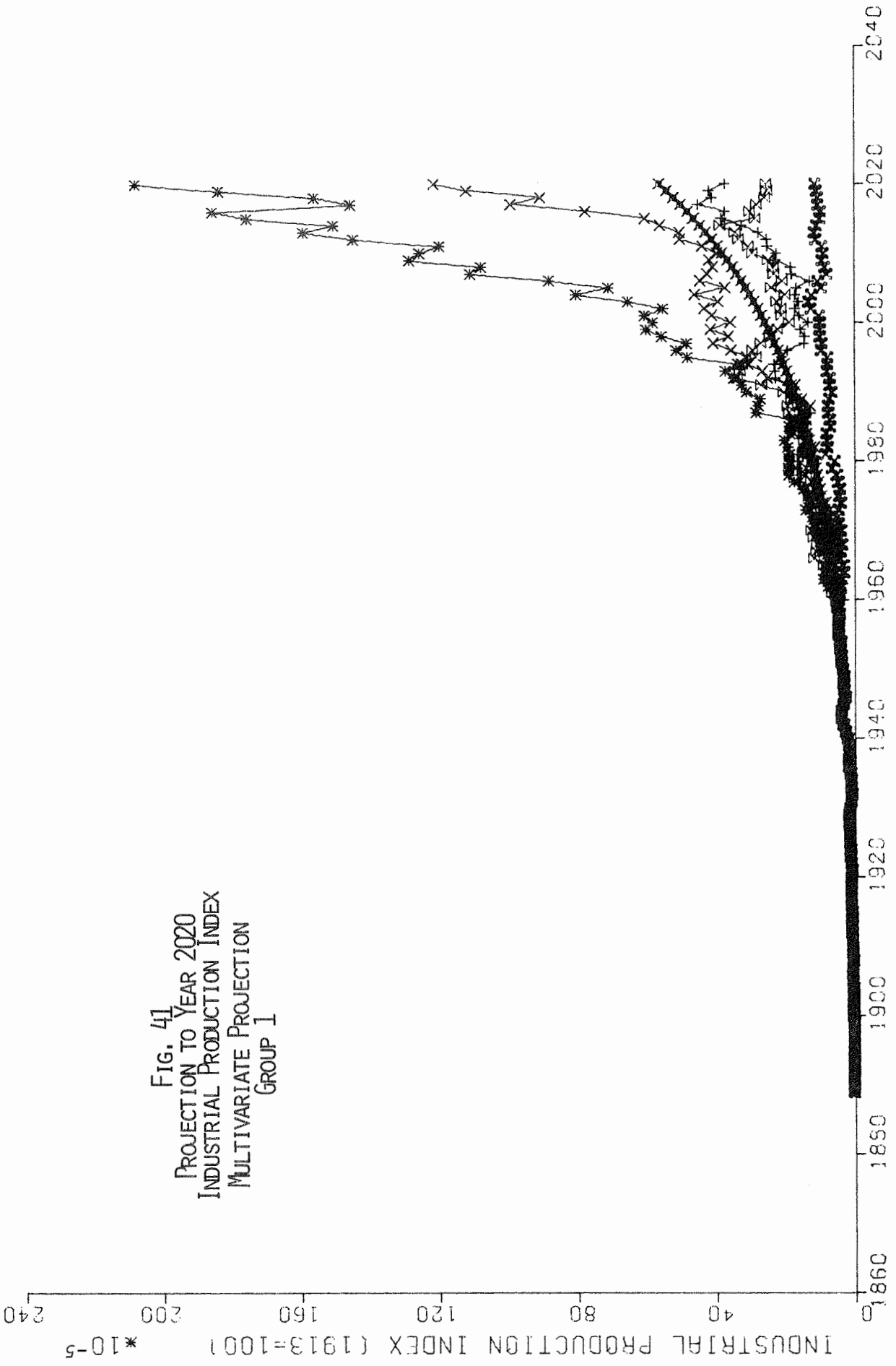


FIG. 41
 PROJECTION TO YEAR 2020
 INDUSTRIAL PRODUCTION INDEX
 MULTIVARIATE PROJECTION
 GROUP 1

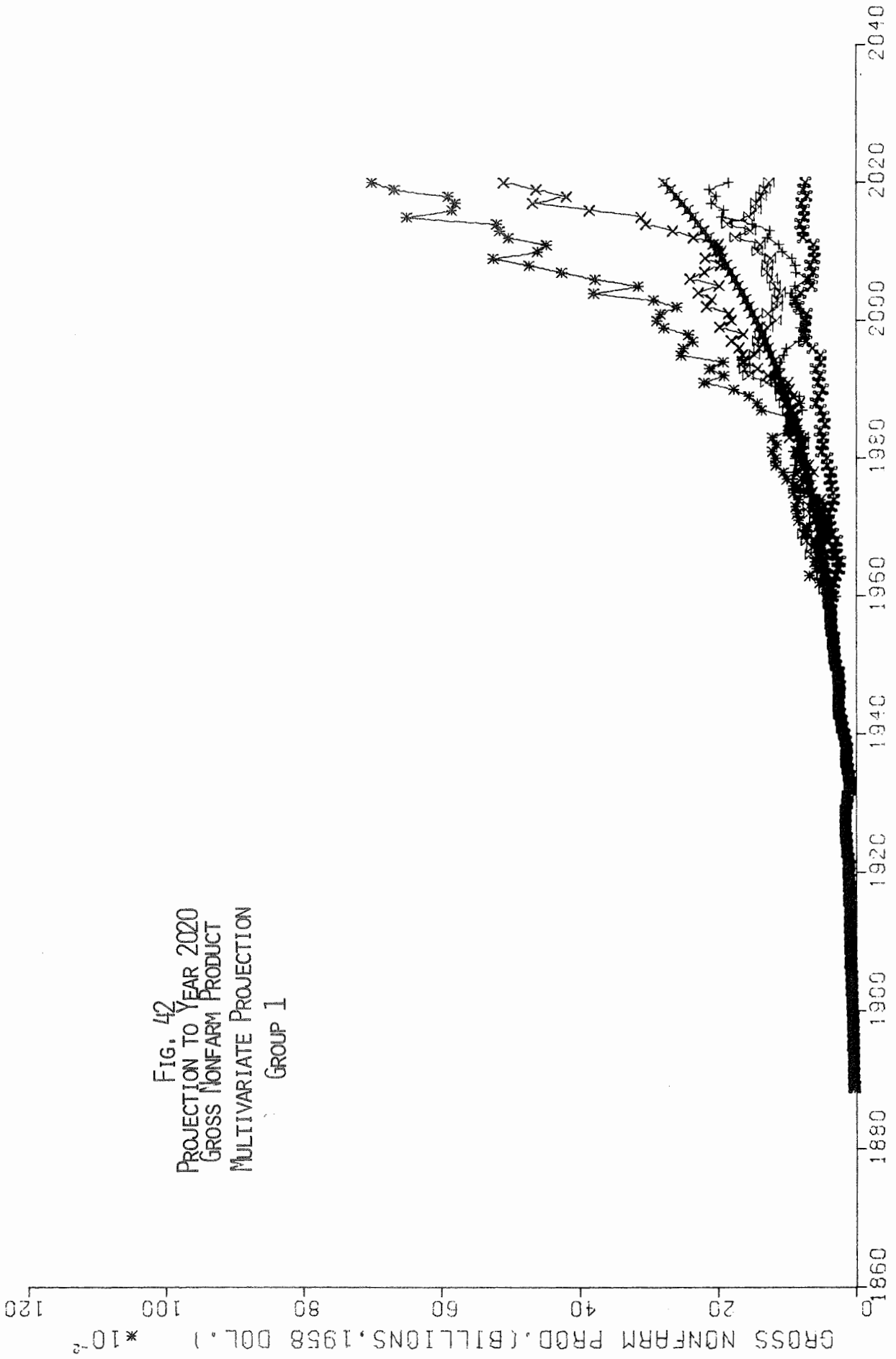


FIG. 42
 PROJECTION TO YEAR 2020
 GROSS NONFARM PRODUCT
 MULTIVARIATE PROJECTION
 GROUP 1

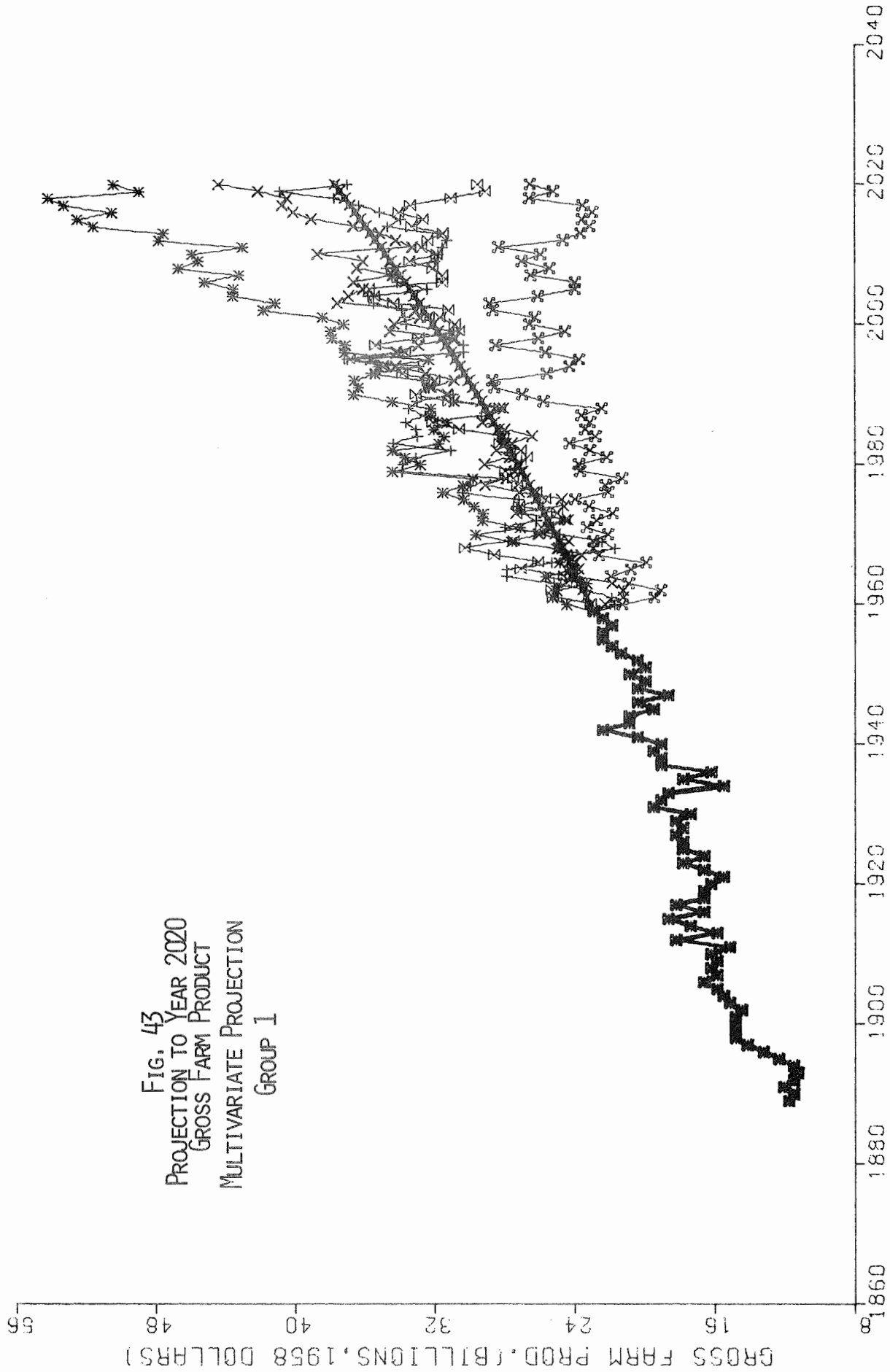


FIG. 43
 PROJECTION TO YEAR 2020
 GROSS FARM PRODUCT
 MULTIVARIATE PROJECTION
 GROUP I

GROSS FARM PROD. (BILLIONS, 1958 DOLLARS)

55

48

40

32

24

16

0

1860

1880

1900

1920

1940

1960

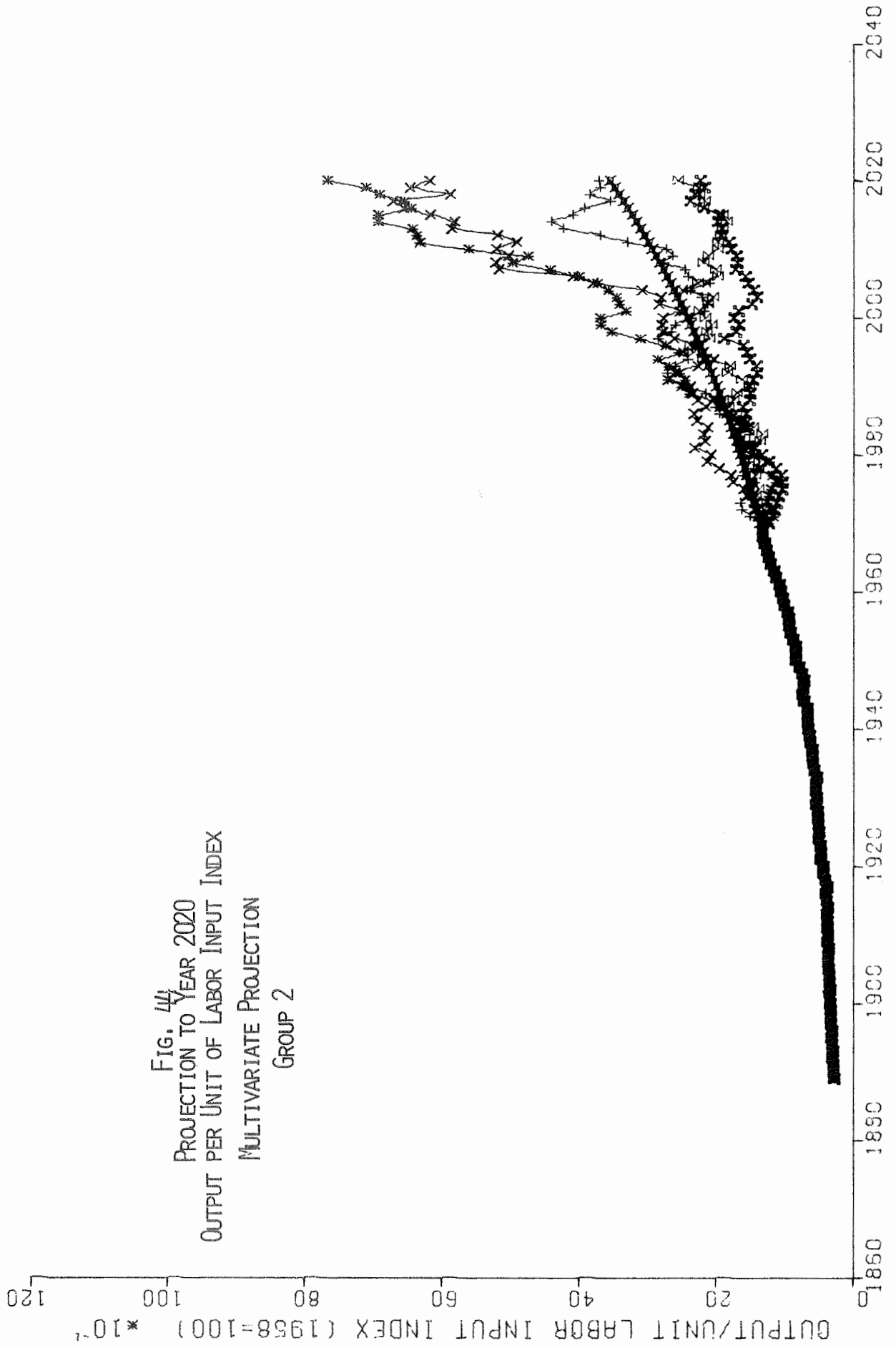
1980

2000

2020

2040

FIG. 44
PROJECTION TO YEAR 2020
OUTPUT PER UNIT OF LABOR INPUT INDEX
MULTIVARIATE PROJECTION
GROUP 2



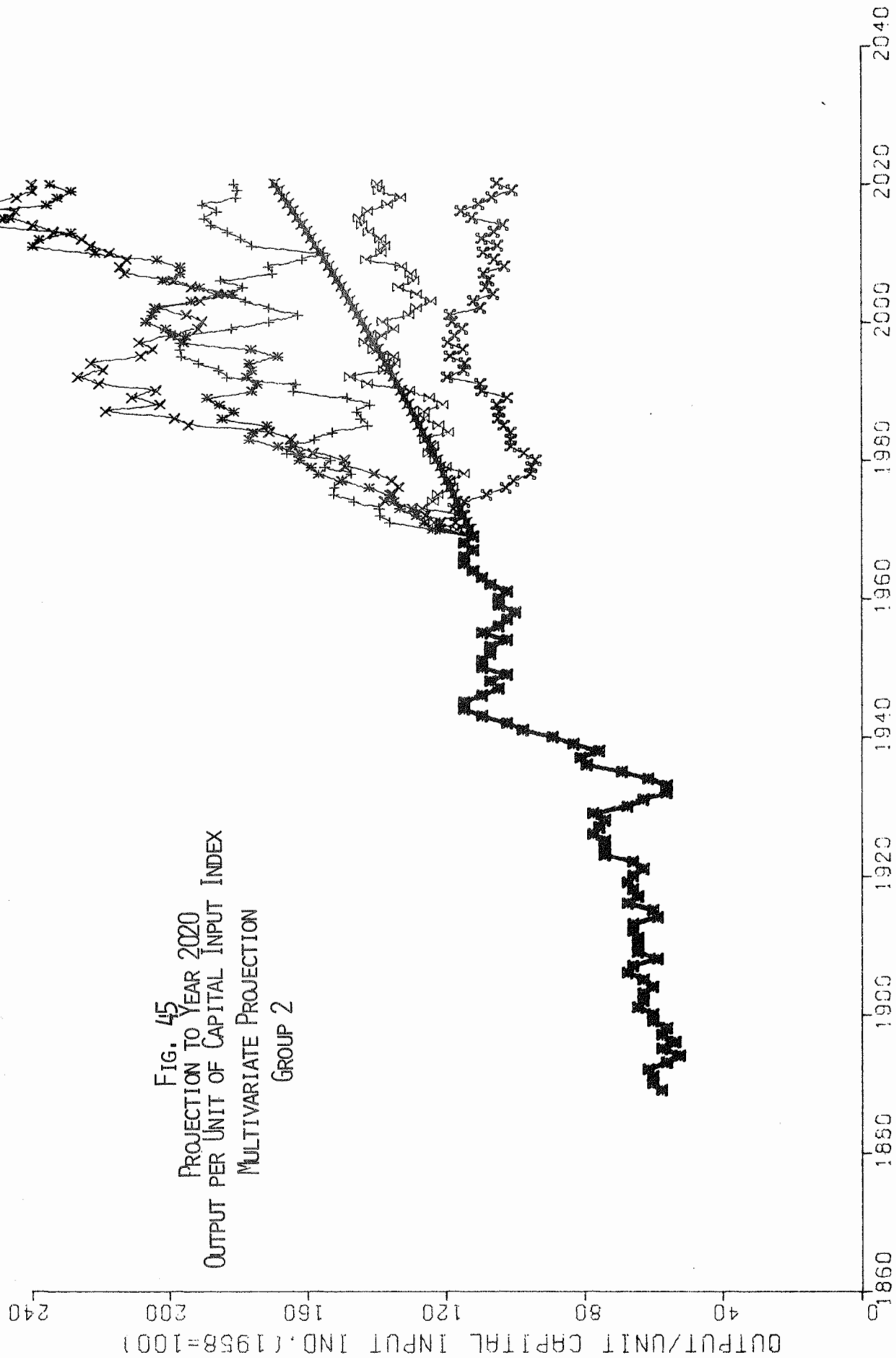


FIG. 45
 PROJECTION TO YEAR 2020
 OUTPUT PER UNIT OF CAPITAL INPUT INDEX
 MULTIVARIATE PROJECTION
 GROUP 2

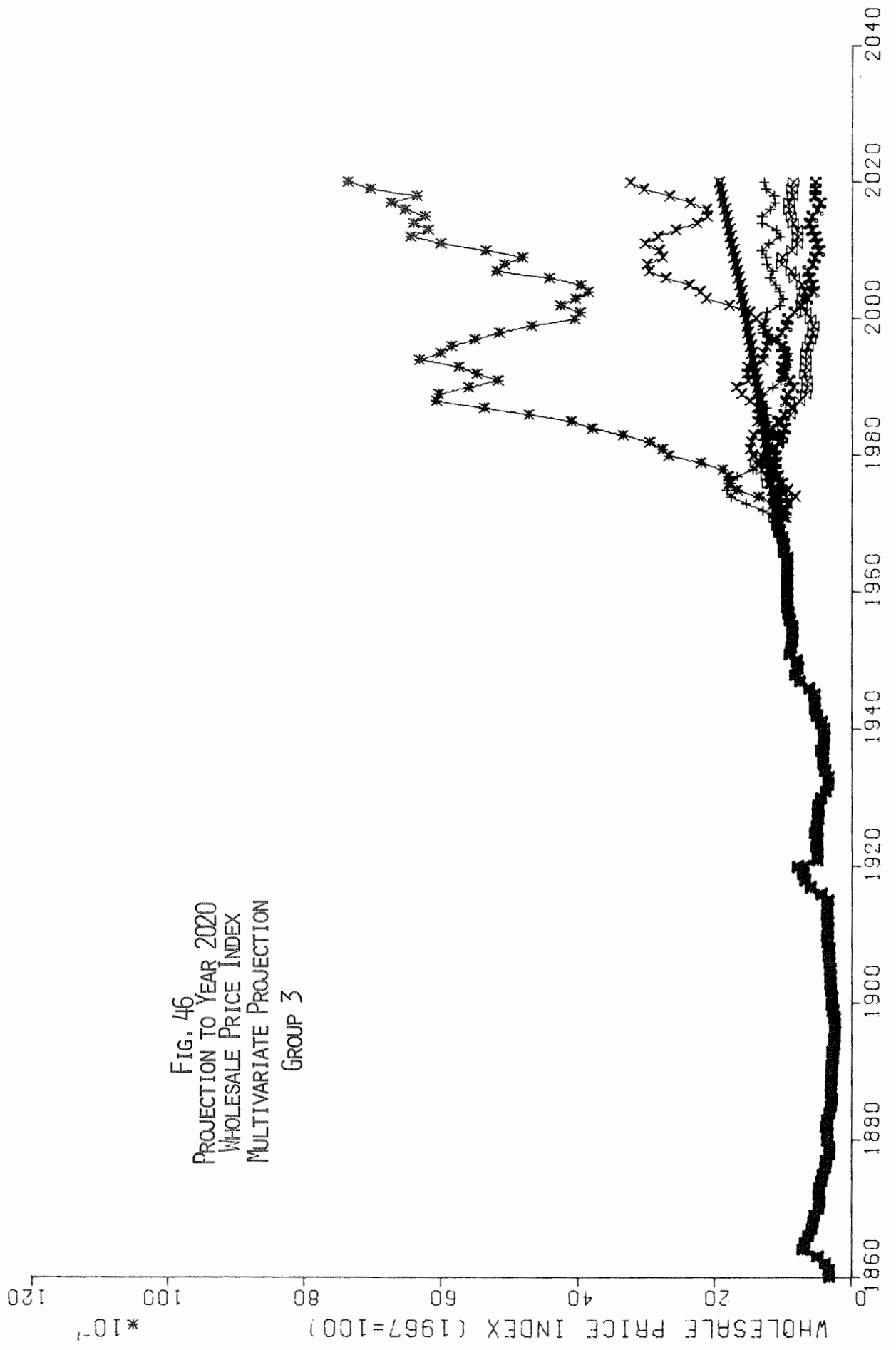


FIG. 46
 PROJECTION TO YEAR 2020
 WHOLESALE PRICE INDEX
 MULTIVARIATE PROJECTION
 GROUP 3

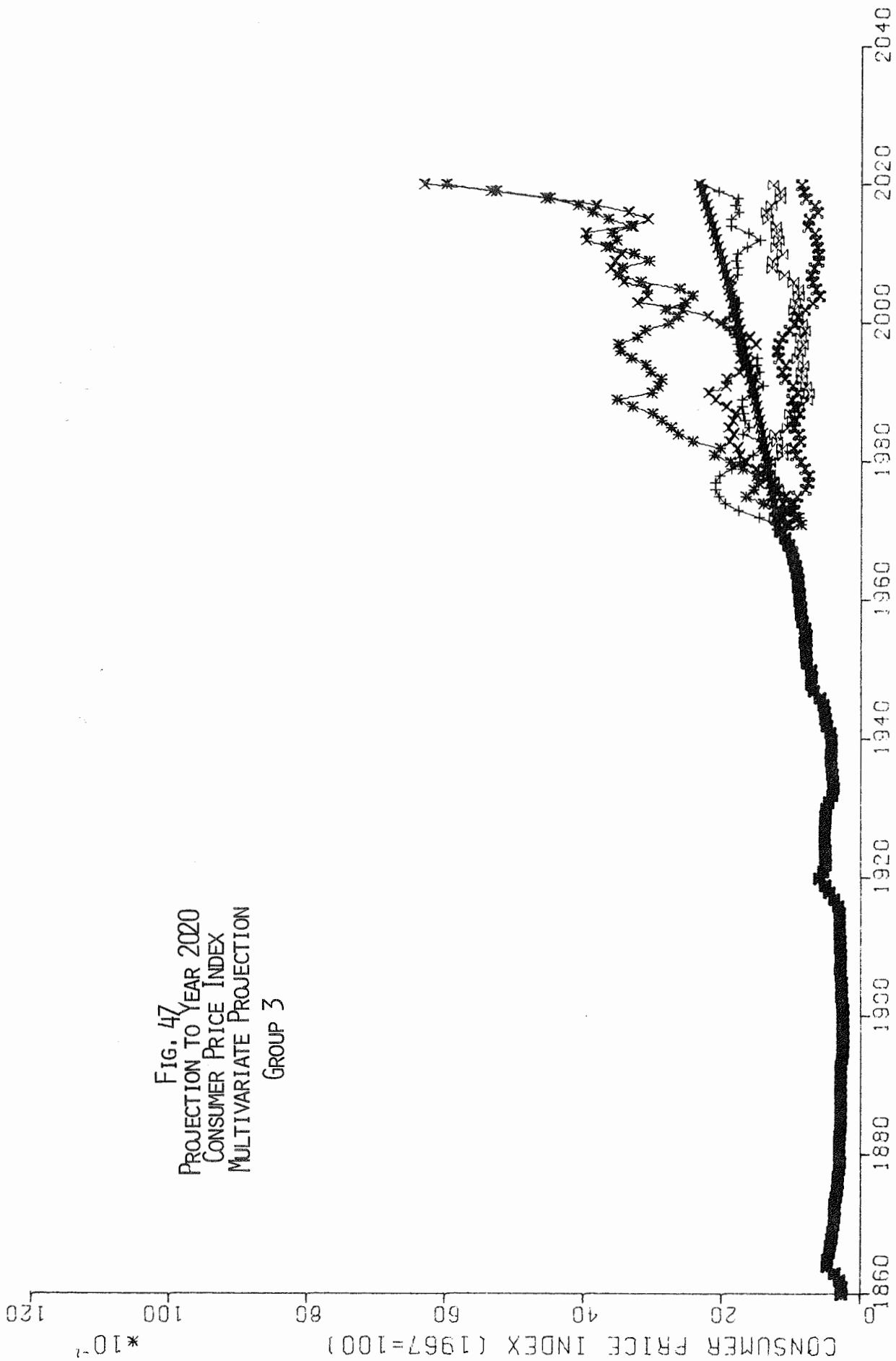


FIG. 47
 PROJECTION TO YEAR 2020
 CONSUMER PRICE INDEX
 MULTIVARIATE PROJECTION
 GROUP 3

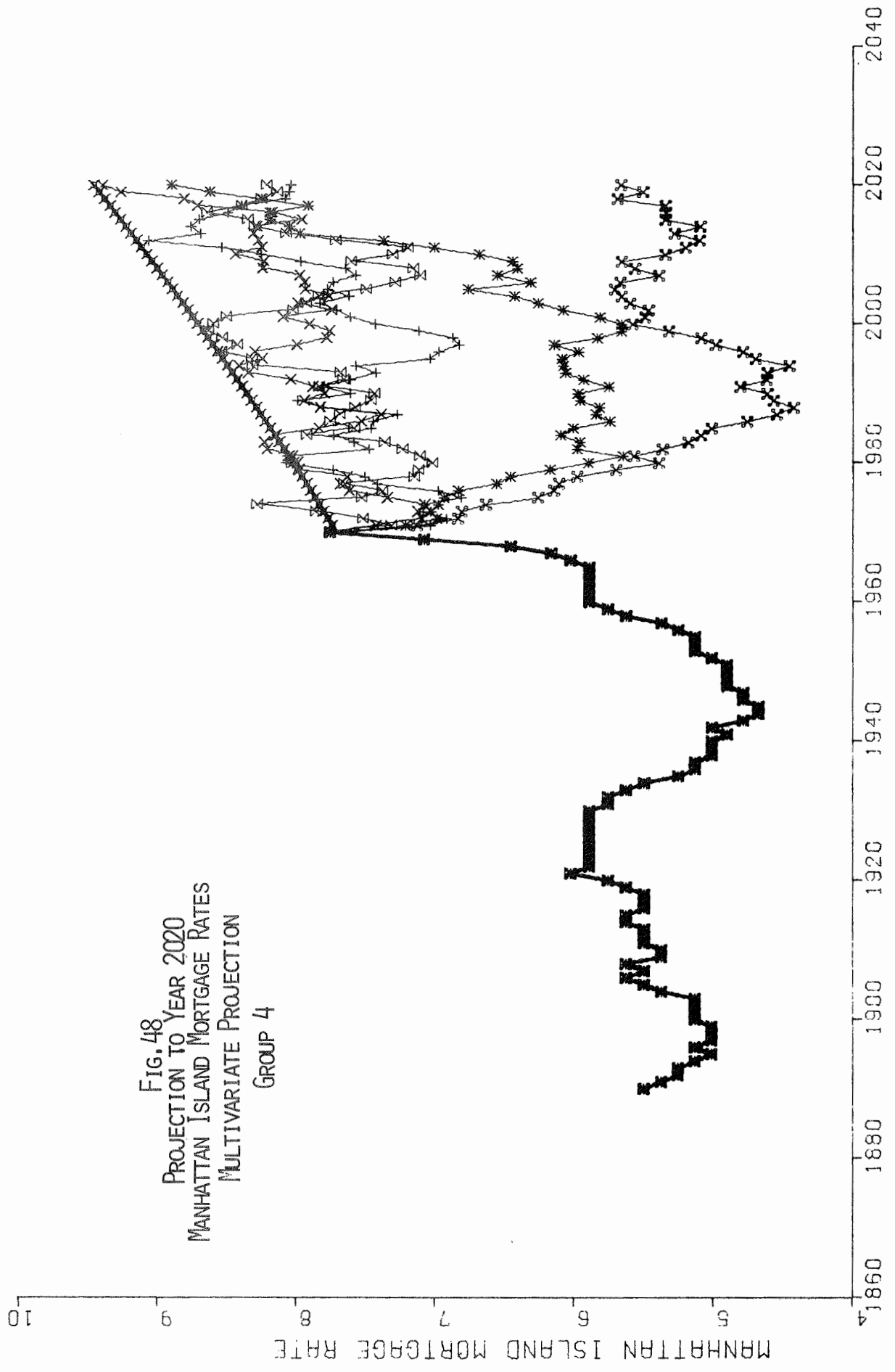


FIG. 48
 PROJECTION TO YEAR 2020
 MANHATTAN ISLAND MORTGAGE RATES
 MULTIVARIATE PROJECTION
 GROUP 4

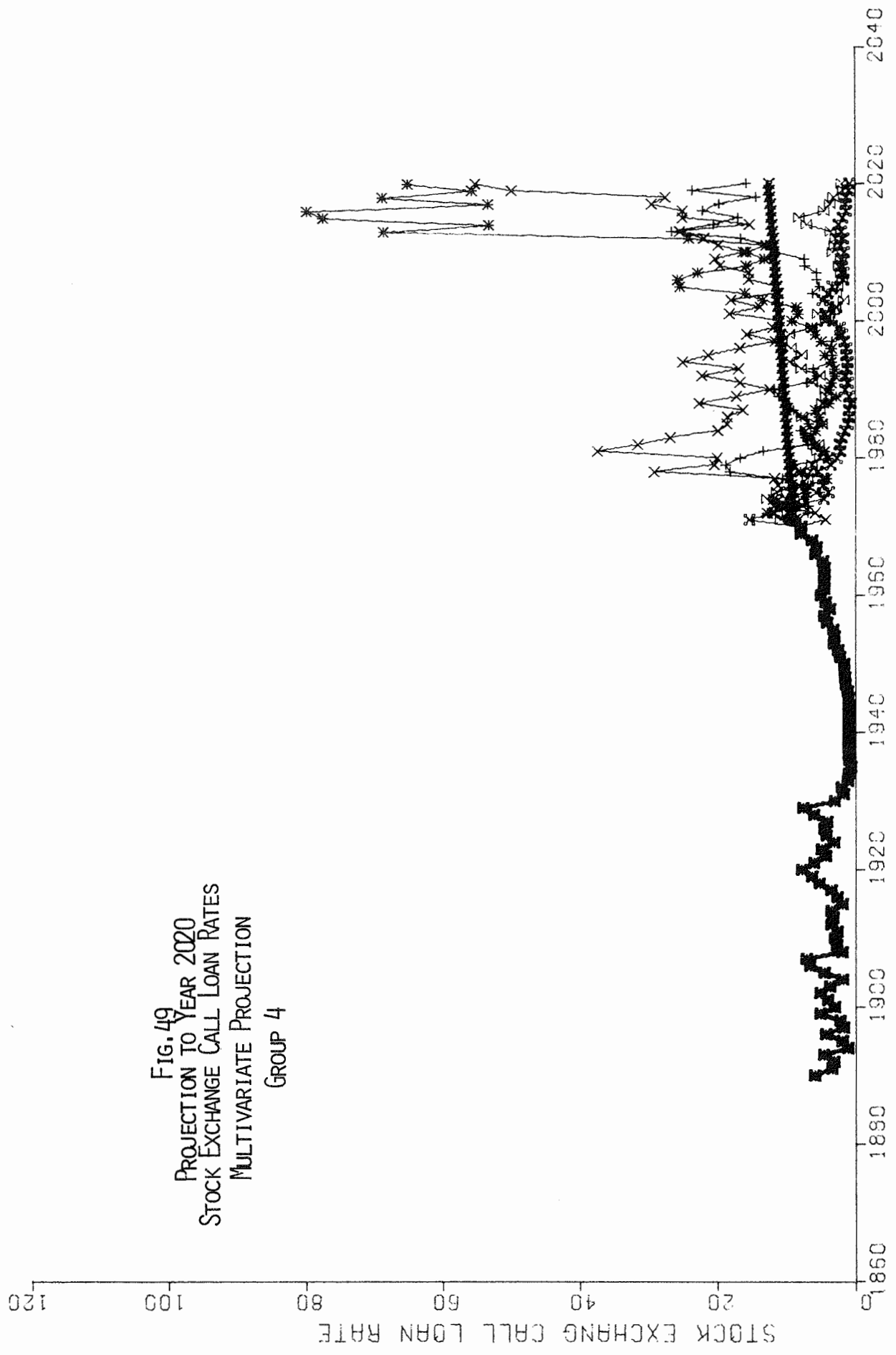
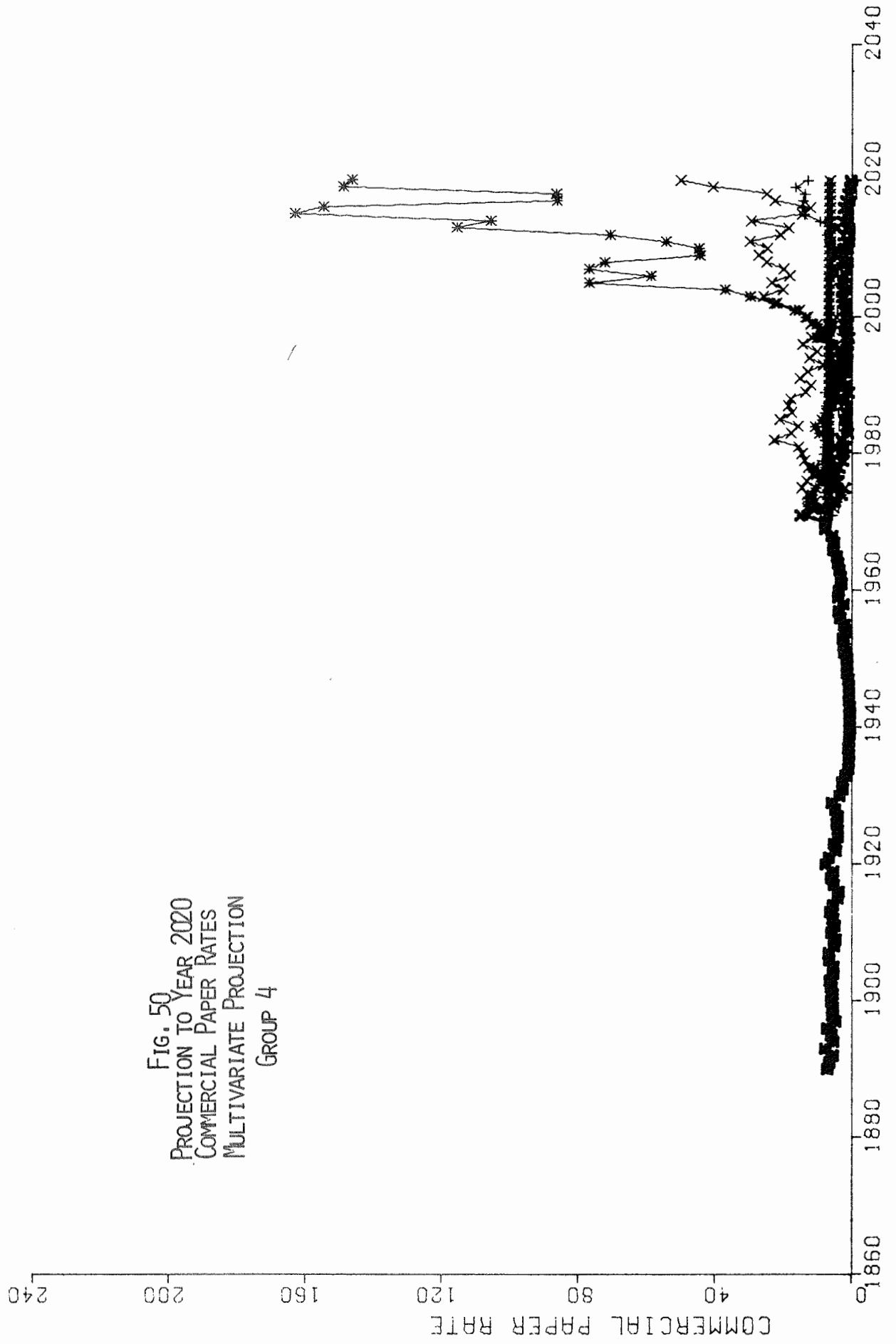


FIG. 49
 PROJECTION TO YEAR 2020
 STOCK EXCHANGE CALL LOAN RATES
 MULTIVARIATE PROJECTION
 GROUP 4

FIG. 50
PROJECTION TO YEAR 2020
COMMERCIAL PAPER RATES
MULTIVARIATE PROJECTION
GROUP 4



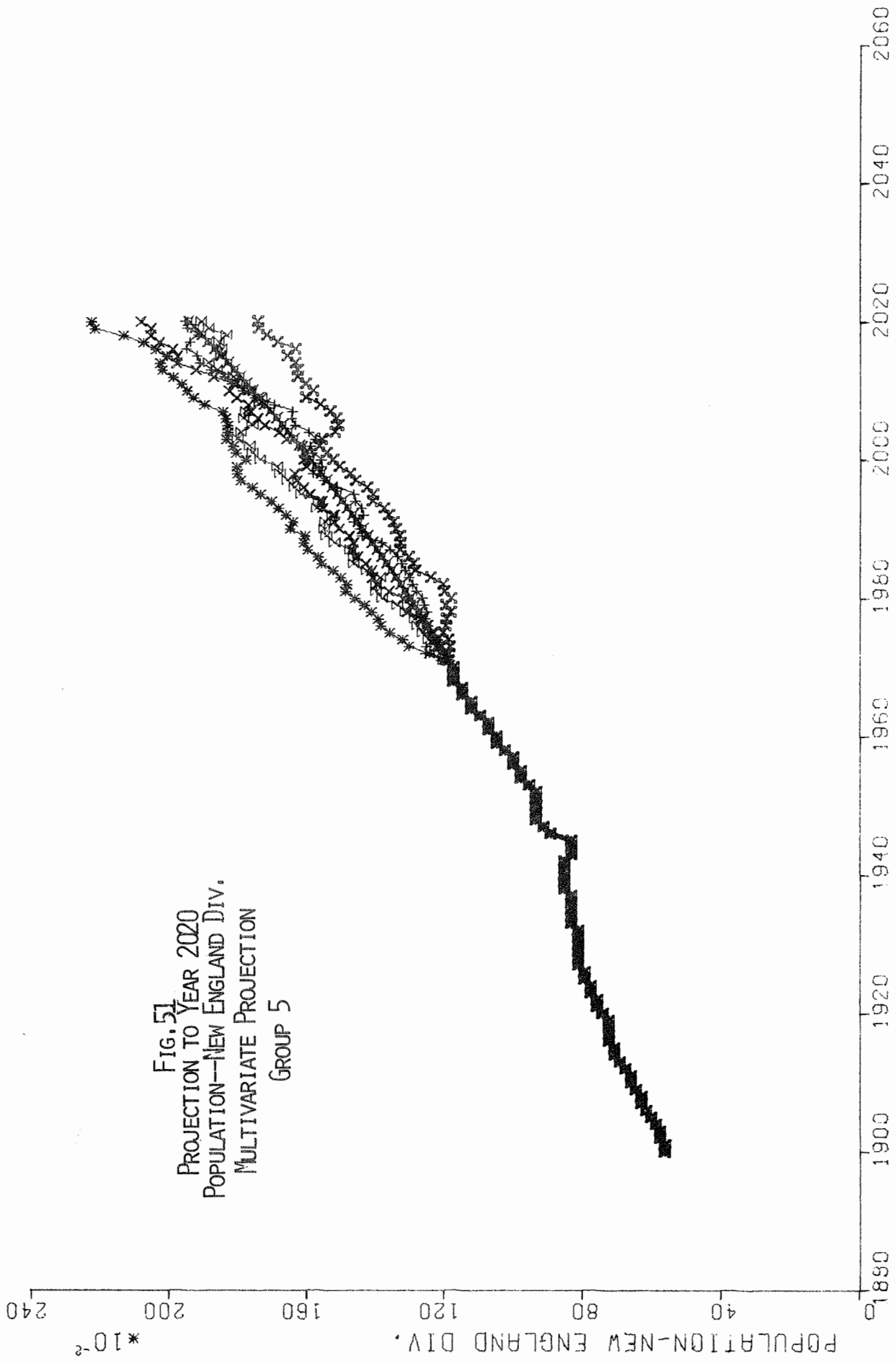


FIG. 51
 PROJECTION TO YEAR 2020
 POPULATION--NEW ENGLAND DIV.
 MULTIVARIATE PROJECTION
 GROUP 5



FIG. 52
 PROJECTION TO YEAR 2020
 POPULATION—MIDDLE ATLANTIC DIV.
 MULTIVARIATE PROJECTION
 GROUP 5

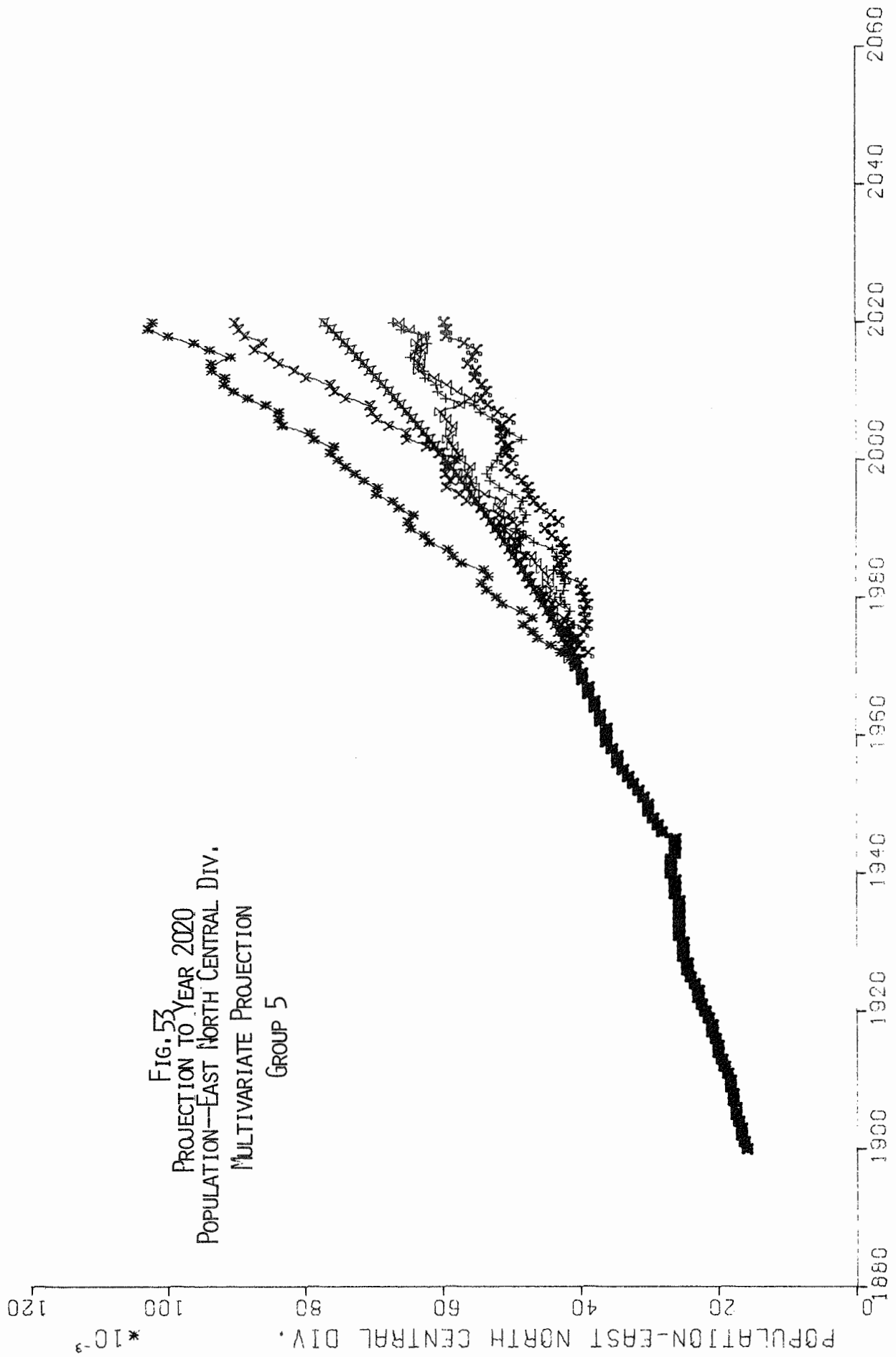


FIG. 53
 PROJECTION TO YEAR 2020
 POPULATION--EAST NORTH CENTRAL DIV.
 MULTIVARIATE PROJECTION
 GROUP 5

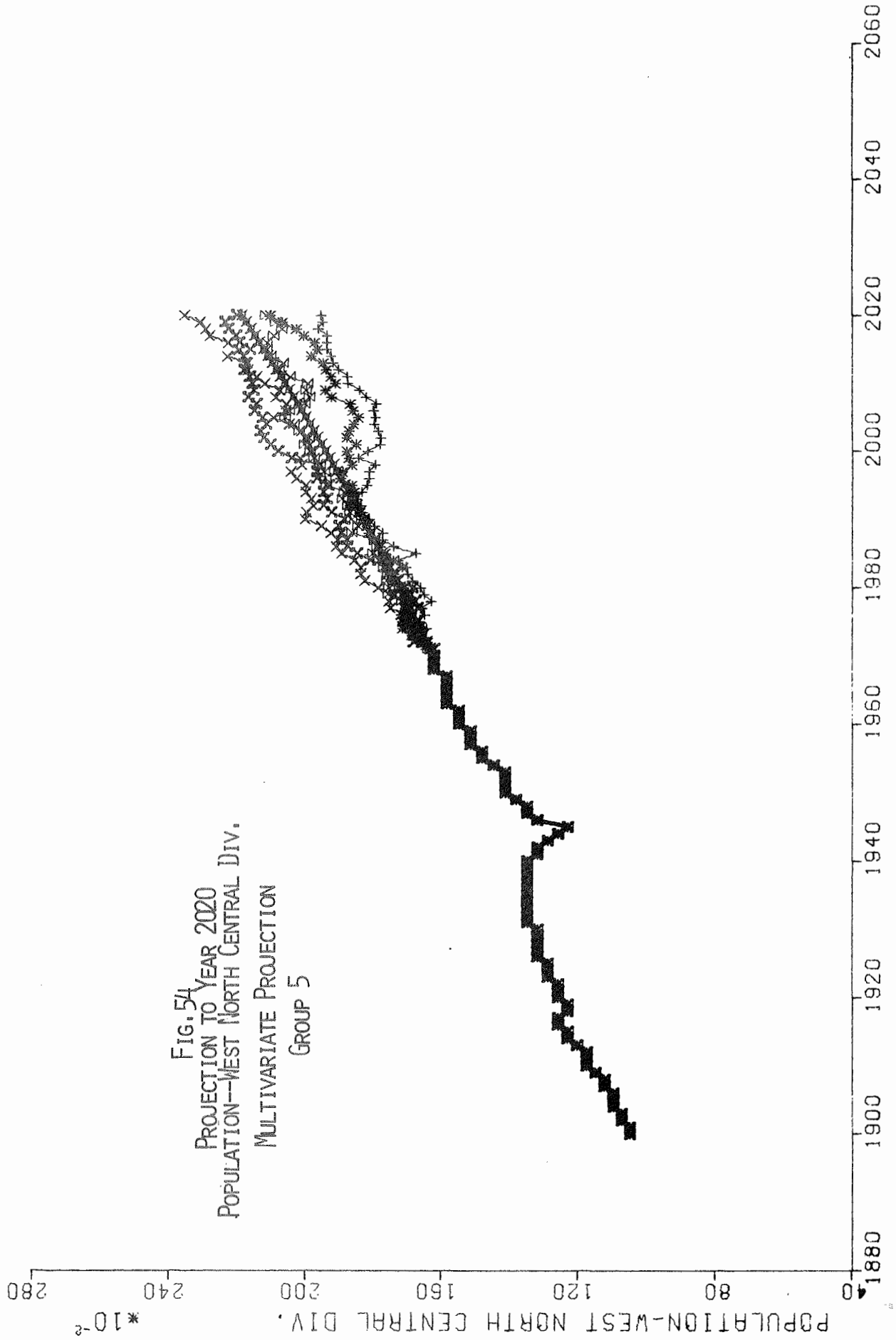


FIG. 54
 PROJECTION TO YEAR 2020
 POPULATION--WEST NORTH CENTRAL DIV.
 MULTIVARIATE PROJECTION
 GROUP 5

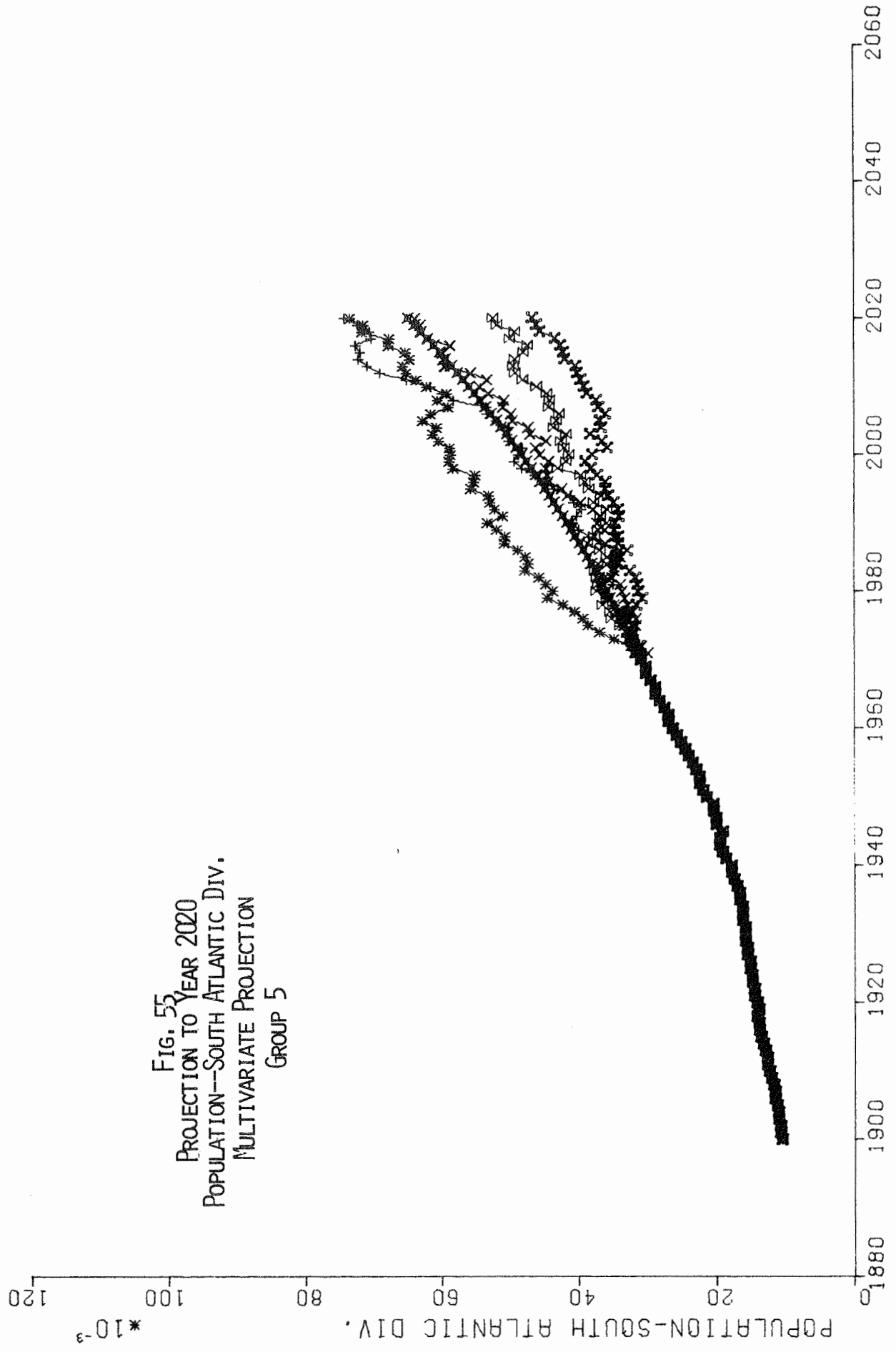


Fig. 55
 PROJECTION TO YEAR 2020
 POPULATION--SOUTH ATLANTIC DIV.
 MULTIVARIATE PROJECTION
 GROUP 5

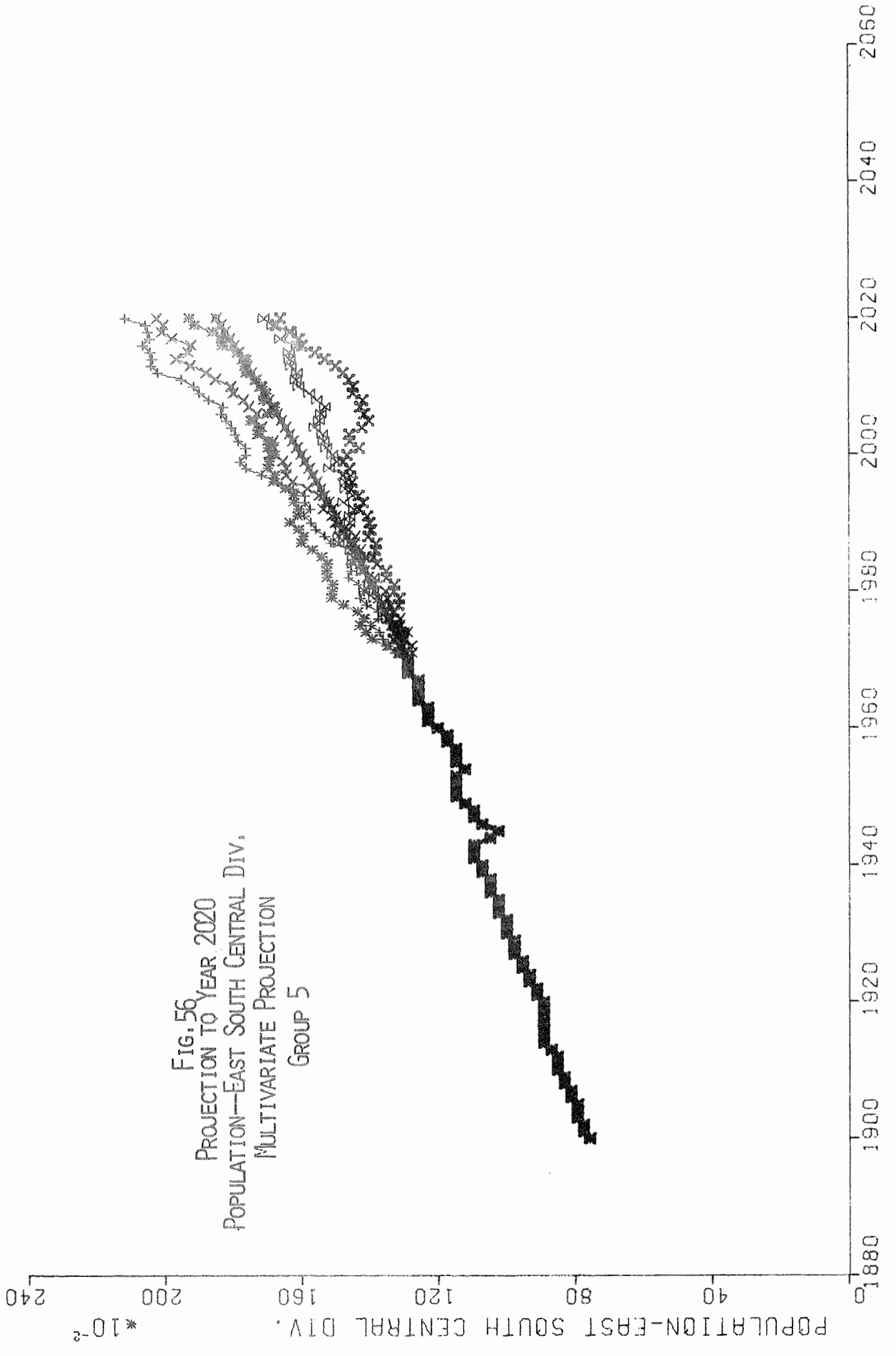


Fig. 56
 PROJECTION TO YEAR 2020
 POPULATION—EAST SOUTH CENTRAL DIV.
 MULTIVARIATE PROJECTION
 GROUP 5

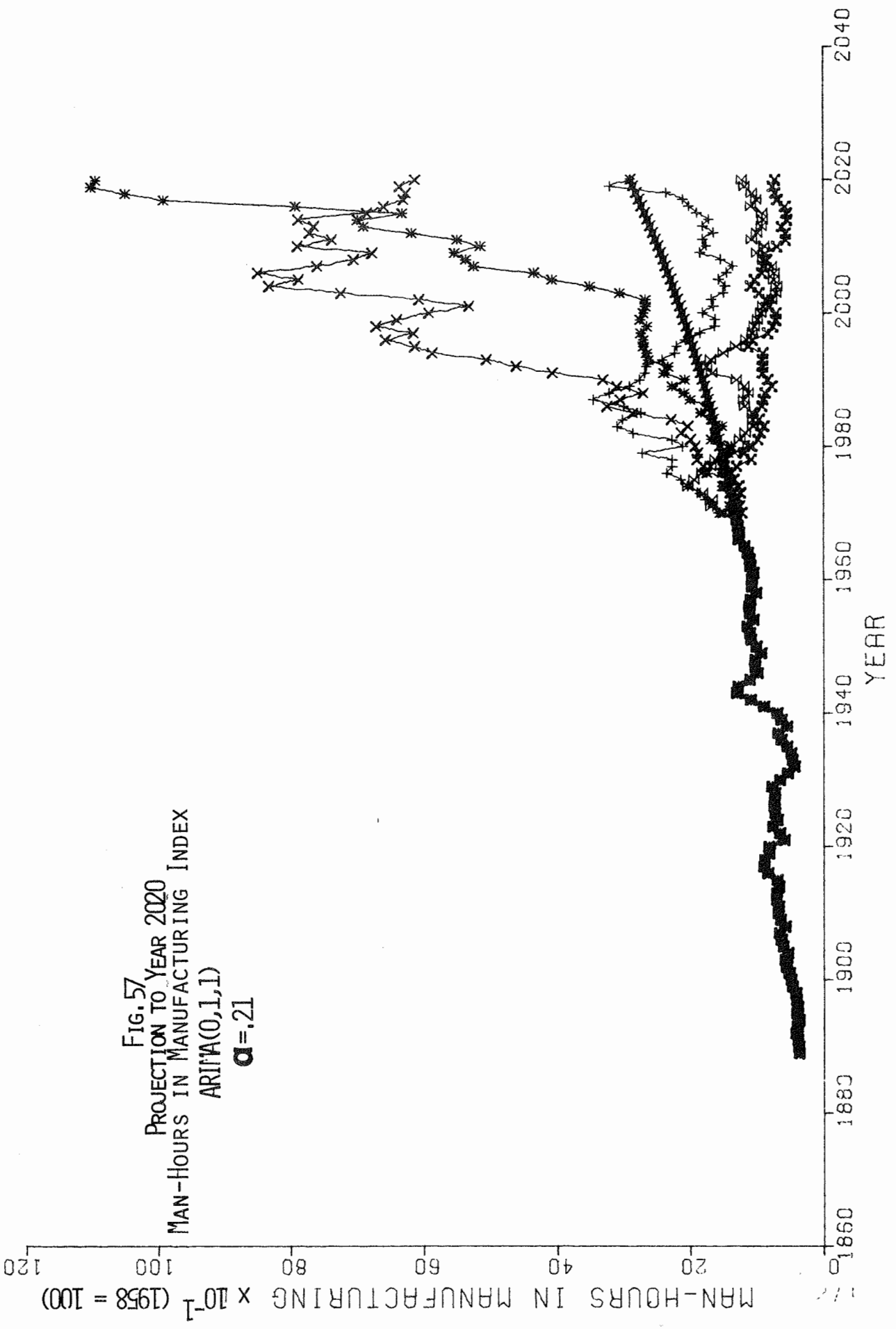


FIG. 57
 PROJECTION TO YEAR 2020
 MAN-HOURS IN MANUFACTURING INDEX
 ARIMA(0,1,1)
 $\alpha = .21$

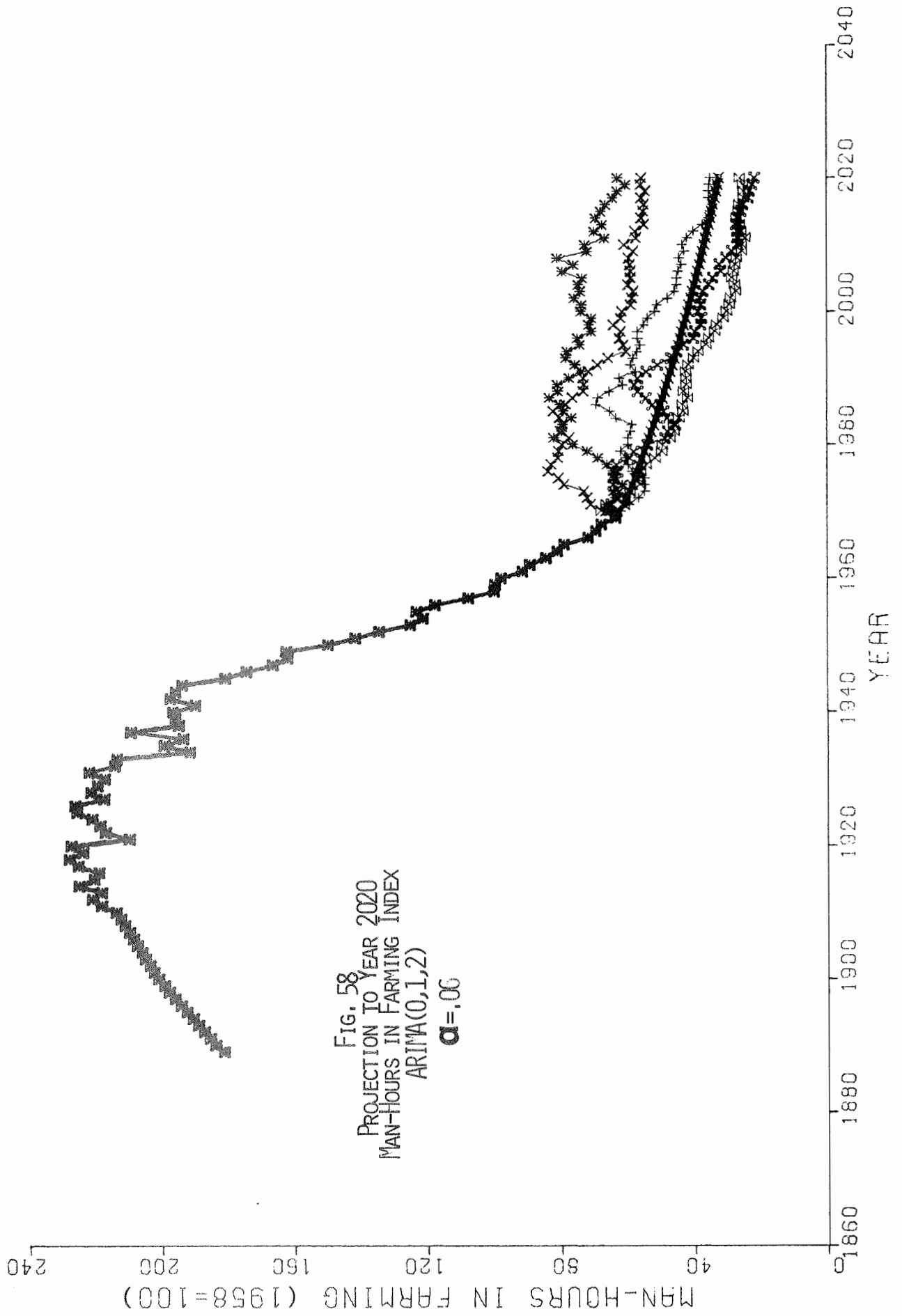


FIG. 58
 PROJECTION TO YEAR 2020
 MAN-HOURS IN FARMING INDEX
 ARIMA(0,1,2)
 $\alpha = .06$

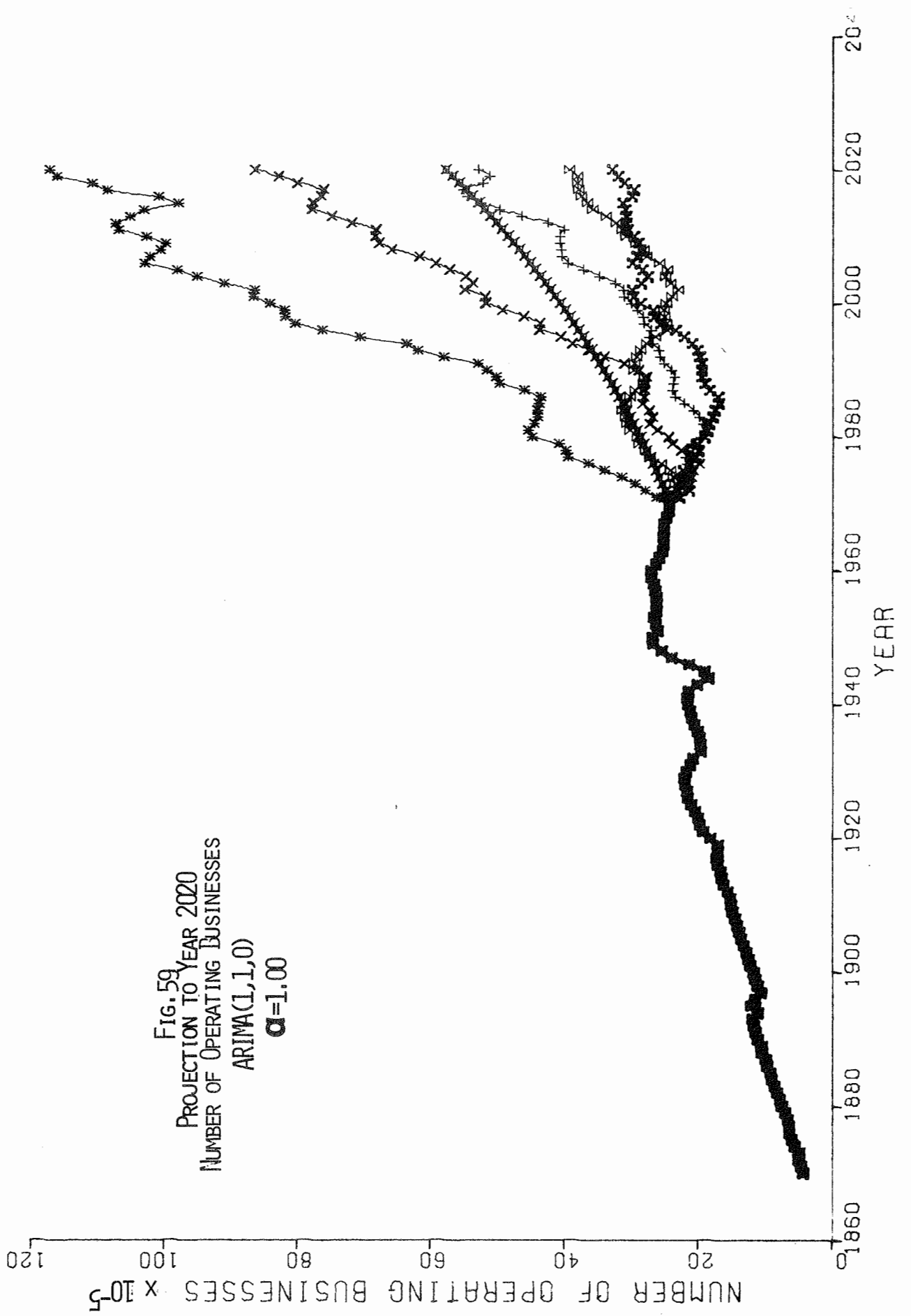


FIG. 59
 PROJECTION TO YEAR 2020
 NUMBER OF OPERATING BUSINESSES
 ARIMA(1,1,0)
 $\alpha = 1.00$

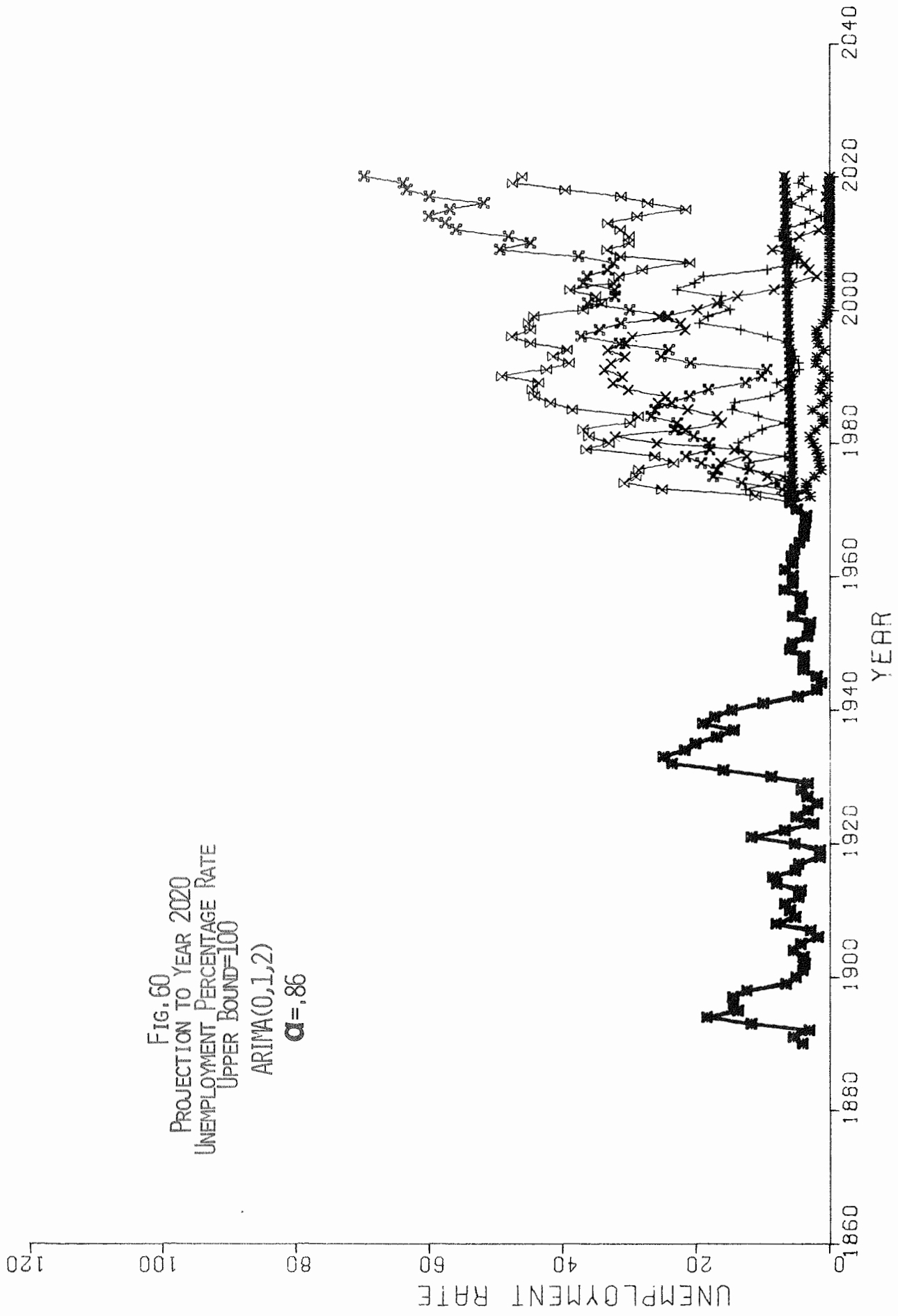


FIG. 60
 PROJECTION TO YEAR 2020
 UNEMPLOYMENT PERCENTAGE RATE
 UPPER BOUND=100
 ARIMA(0,1,2)
 $\alpha = .86$

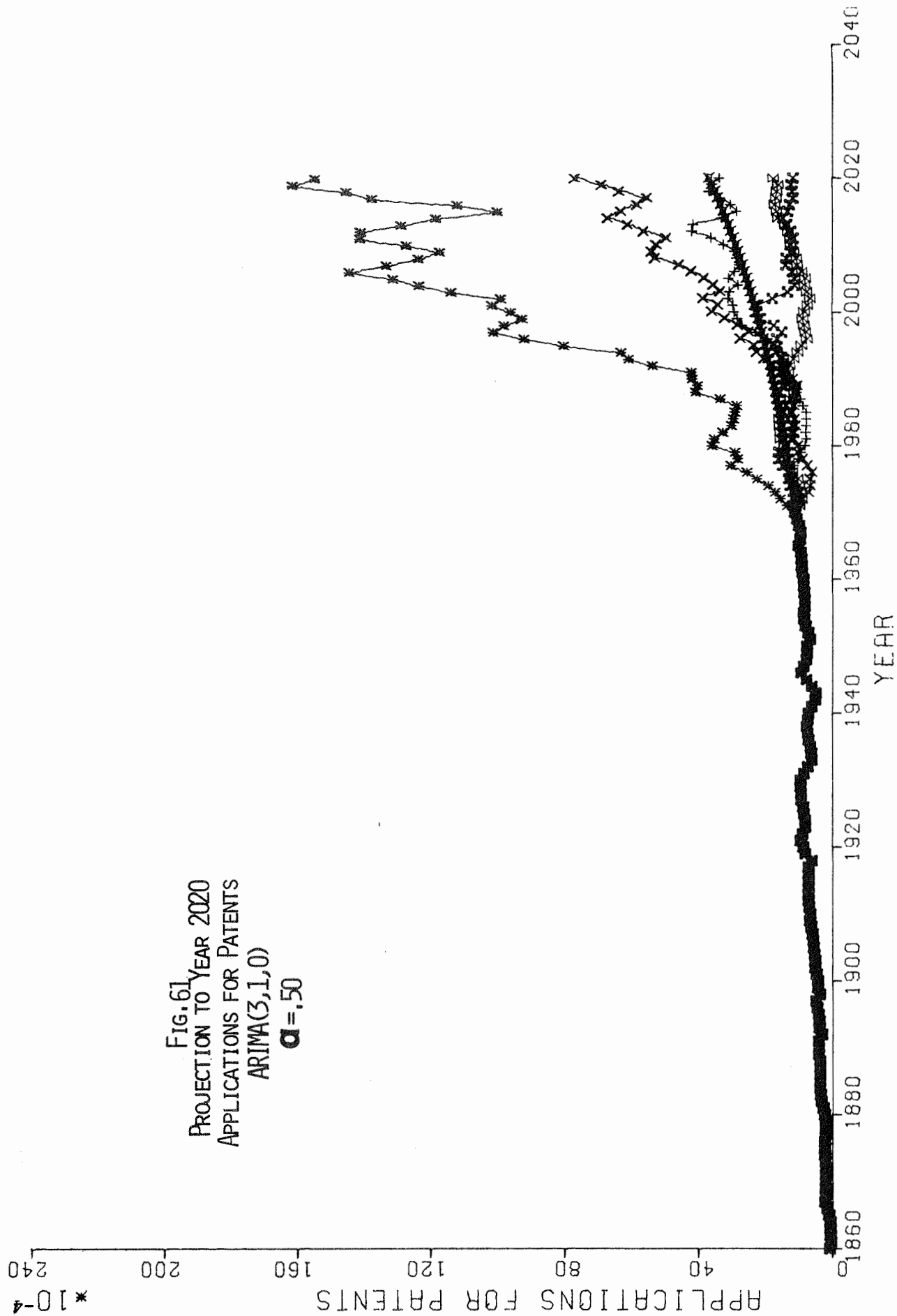


FIG. 61
 PROJECTION TO YEAR 2020
 APPLICATIONS FOR PATENTS
 ARIMA(3,1,0)
 $\alpha = .50$

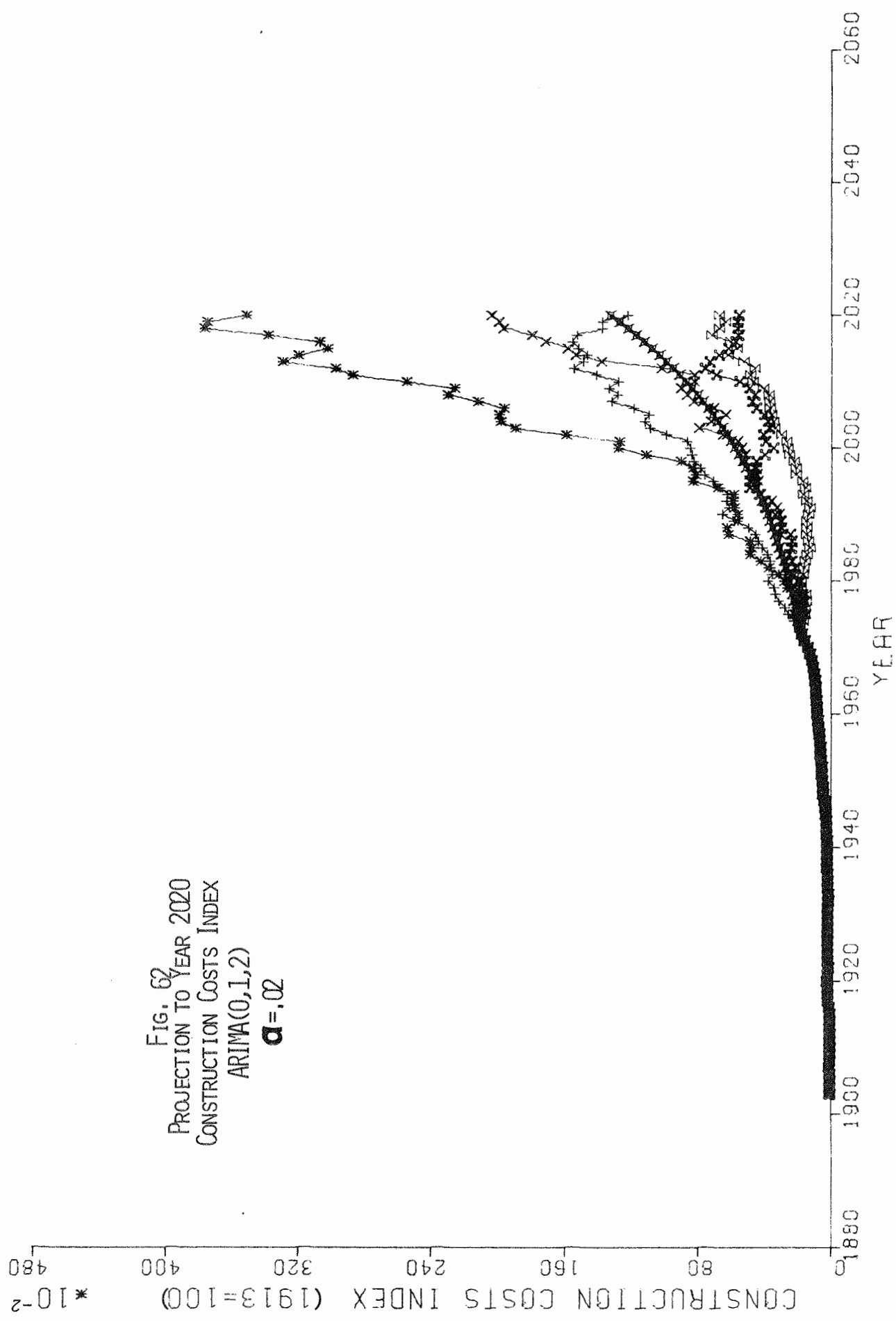


FIG. 62
 PROJECTION TO YEAR 2020
 CONSTRUCTION COSTS INDEX
 ARIMA(0,1,2)
 $\alpha = .02$

APPENDIX A
DESCRIPTION OF COMPUTER PROGRAM SVPROJ

INTRODUCTION

1. Origin of Program. This program was developed by Shin Chang of the Center for Research in Water Resources, the University of Texas at Austin under the supervision of Leo R. Beard, Technical Director. Subroutines obtained from other sources are:

- a. FTARPS, FTMAPS, FTAUTO, FTCAST, FTGENI, MDCH and their supporting subroutines FTFUNC, GGNOR, GGUI, LEQTIF, LUDATF, LUELMP, MERF, MERFI, MGAMMA, UERTST, VABMXF and ZSYSTEM. All are included in the International Mathematics and Statistical Library (IMSL) computer package.
- b. PLOT was developed by Mr. David Ford and Mr. David Lott of the Center for Research in Water Resources, The University of Texas at Austin.

2. Capability of the SVPROJ Program. SVPROJ is capable of performing the following jobs:

- a. Identification of the degree of differencing, d , and number of parameters p and q of the best-fit ARIMA (p,d,q) model for time series of a single variable.
- b. Estimation of parameters of an identified ARIMA (p,d,q) model.
- c. Generation of 51 independent sets of projections using the best-fit ARIMA (p,d,q) model for a single variable.
- d. Plotting of up to 7 selected projections on one chart.

3. Hardware and Software Requirements. This program has been developed and tested on the CDC 64/6600 computer system. No tapes or disks are required. The Fortran IV compiler and the IMSL library are used.

DESCRIPTION OF PROGRAM

4. Program Organization. The SVPROJ program consists of a main program and 23 subroutines, of which 18 are subroutines from IMSL. All input

data and input control parameters are read in the main program from punched cards. The program will make a normal exit (unless errors occur) only in the main program.

5. Method of Computation. The computational procedures and equations used in this program are described in chapter 4 of the main report under the title "Use of the ARIMA Model."

6. Description of Subroutines and Variables. The function of each subroutine and the definitions of variables in each subroutine are given in the program listings.

PROGRAM USE

7. Dimensional Limitations of the Program. For practical purpose, the following limitations are set in the program:

<u>Variable</u>	<u>Size</u>	<u>Definition</u>
LSPLIT	150	length of calibration series
LPROJ	100	length of projection
NSET	51	Sets of independent projections

These dimensions can be easily extended by increasing the dimension specification in related arrays, however, ultimate limitations are determined by the available central memory of the computer system used.

8. Control Parameters for Optional Jobs. The following parameters are used to control the execution of certain optional jobs in this program.

IBOND:	Controls the execution of lower and/or upper bound on the values of projections
ISPLIT:	Controls the execution of the split-record
ISPY:	Controls the execution of the model parameter identification for the calibration series
LOGX:	Controls the execution of natural logarithmic transformation of the calibration series.
NPRINT:	Controls the execution of the printed output of the selected projections (7 sets)

NPUNCH: Controls the execution of the punched-card output of the selected projections (7 sets).

NPLOT: Controls the execution of the plotted output of the selected projections (7 sets sets, or 5 sets).

9. Preparation Of Input Variables.

<u>Variable</u>		<u>Description</u>
IBOND: (Requires LOGX=1)	-1	a nonzero lower bound control is used
	0	a zero lower bound is used
	1	a zero lower bound and a nonzero upper bound is used
	2	both nonzero lower and upper bounds are used
ID	integer	degree of differencing, d, in an ARIMA (p,d,q) model
IP	integer	number of autoregressive parameters, p, in an ARIMA (p,d,q) model
IQ	integer	number of moving average parameters, q, in an ARIMA (p,d,q) model
ISPLIT	1	split-record test of the calibrated-ARIMA model is desired,
	0	not desired
ISPY	1	identification of ID, IP, and IQ for an ARIMA model by the program is desired
	0	ID, IP, IQ are provided by the user through input.
IYRB		beginning year of record of the calibration series
IYRE		ending year of record of the calibration series
IYRP		ending year of record of the projection series
LOGX	1	natural logarithmic transformation of the variable is desired,
NPLOT	0	no plot is desired as output
	1	plots of data and 7 projections depicted in one chart are desired
	2	plots of data and 7 projections, and data and 5 projections, depicted in 2 charts are desired

	3	plots of data and 5 projections depicted in one chart are desired
NPRINT	1	7 sets of projections of the variable being analyzed will be printed as output
	0	not desired
NPUNCH	1	7 sets of projections of the variable being analyzed will be punched as output
	0	not desired
NV	integer	number of variables to be analyzed sequentially in a multiple job run
YLABEL (8)	character	name of the variable
Z (150)	real	a vector of N elements containing data of the calibration series, and N is computed internally as IYRE-IYRB + 1
ZLO	real	lower bound of the variable being analyzed
ZUP	real	upper bound of the variable being analyzed

10. Preparation of Input Cards. The following eight types of input cards are required.

<u>Card</u>	<u>Format</u>	<u>Variable(s)</u>
A	16I5	NV
B	8A10	YLABEL
C	16I5	IYRB, IYRE, IYRP
D	8A10	IFMT
E	IFMT	Z
F	16I5	ISPLIT, LOGX, ISPY, IBOND, NPRINT, NPUNCH, NPLOT
G	16I5	IP, ID, IQ
H	2F20.4	ZUP, ZLO

It should be noted that the G card is not required if ISPY = 1, and the H card is not required if LOGX = 0 or IBOND = 0.

11. Arrangement of Input Cards.

a. Arrangement of the input deck for a single job is:

A, B, C, D, E, E...E, F, G, H

b. Arrangement of the input deck for a multiple job (maximum 14 jobs)

A, B, C, D, E, ...E, F, G, H, B, C, D, E, ...E, F, G, H, ...

1st job
2nd job
3rd job

12. Overview of Output. The following printed output is shown for each job:

- a. IP, ID, IQ of the identified ARIMA (p,d,q) model.
- b. α , the significance level of goodness of fit
- c. estimated parameters (θ_0 , Ψ_i and θ_i) of the ARIMA (p,d,q) model

Printed and punched values of 7 sets of projections, and their plots in one curve are optional outputs.

13. Program Listing. Source program listing for SVPROJ and all its subroutines are shown in the following pages.


```

1864 DO 1118 JPC=1,NV
1874 REWIND NTAPE
1884 DO 1078 JT=1,N
1894 JS=INDEX(J)
1904 CONTINUE
1914 CALL FOCASZ (Y,N,JPC)
1924 DO 1088 JS=1,NSET
1934 U(JS,JPC)=SIM(LPROJ,JS)
1944 CONTINUE
1954 DO 1098 JT=1,LPROJ
1964 D(JT,MP1,JPC)=FCST(2,JT)
1974 CONTINUE
1984 CC STORE ALL SIMULATIONS OF P-C IN DISK
1994 NTAPE=JPC+12
2004 DO 1108 JS=1,NSET
2014 WRITE (NTAPE,1430) (SIM(I,JS),I=1,LPROJ)
2024 CONTINUE
2034 1108 CONTINUE
2044 CC CONSTRUCT LAST MEMBERS OF SIMULATED VARIABLE USING LAST
2054 MEMBERS OF SIMULATED P-C
2064 DO 1144 JVC=1,NV
2074 DO 1154 JS=1,NSET
2084 TEMP=0.0
2094 DO 1124 JPC=1,NV
2104 TEMP=TEMP+U(JS,JPC)*Z(JV,JPC)
2114 CONTINUE
2124 TEMP=TEMP*SDV(JV)*XMEAN(JV)
2134 IF (LOGX.EQ.0) TEMP=10.0**TEMP
2144 ZL(JS,JV)=TEMP
2154 1124 CONTINUE
2164 1154 CONTINUE
2174 CC COMPUTE WEIGHTED SUM OF LAST MEMBERS OF SIMULATED VARIABLE
2184 DO 1164 JS=1,NSET
2194 TEMP=0.0
2204 DO 1174 JVC=1,NV
2214 TEMP=TEMP+WT(JV)*ZL(JS,JVC)
2224 CONTINUE
2234 WT=SUM(JS)*TEMP
2244 1164 CONTINUE
2254 CC A=WT*SUM(I)
2264 RANK WTSUM( ) FROM MAX TO MIN
2274 DO 1174 JS=1,NSET
2284 IF (WTSUM(JS).LT.A) A=WT*SUM(JS)
2294 CONTINUE
2304 A=10.0
2314 DO 1184 JS=1,NSET
2324 A=AMIN(A,WTSUM(JS))
2334 IF (WTSUM(JS).LT.A) GO TO 1180
2344 1174 CONTINUE
2354 ID=JS
2364 WTSUM(CID)=AMIN
2374 INDEX(CID)=ID
2384 1184 CONTINUE
2394 CC CONSTRUCT SELECTED REPRESENTATIVES OF NBETS SIMULATED P-C
2404 DO 1238 JPC=1,NV
2414 NTAPE=JPC+12
2424 REWIND NTAPE
2434 DO 1208 JS=1,NSET
2444 READ (NTAPE,1430) (SIM(I,JS),I=1,LPROJ)
2454 1208 CONTINUE
2464 1238 CONTINUE
2474 1864 DO 1224 I=1,MSET
2484 J=MODE(I)
2494 JS=INDEX(J)
2504 DO 1218 JT=1,LPROJ
2514 D(JT,I,JPC)=SIM(JT,JS)
2524 CONTINUE
2534 1218 CONTINUE
2544 1224 CONTINUE
2554 1238 CONTINUE
2564 CC
2574 CC FORECASTS AND SIMULATIONS VIA THE PROJECTED P-C
2584 DO 1414 JVC=1,NV
2594 DO 1278 JT=1,LPROJ
2604 JJ=JT+L9PLIT
2614 DO 1254 JS=1,MSET
2624 TEMP=0.0
2634 DO 1244 JPC=1,NV
2644 TEMP=TEMP+D(JT,JS,JPC)*Z(JV,JPC)
2654 CONTINUE
2664 TEMP=TEMP*SDV(JV)*XMEAN(JV)
2674 U(CJJ,JS+1)=TEMP
2684 1254 CONTINUE
2694 U(CJJ,I)=0.0*(U(CJJ,2)-U(CJJ,MP1))
2704 U(CJJ,MP2)=2.0*(U(CJJ,MP1)-U(CJJ,2))
2714 TEMP=0.0
2724 DO 1264 JPC=1,NV
2734 TEMP=TEMP+D(JT,MP1,JPC)*Z(JV,JPC)
2744 CONTINUE
2754 TEMP=TEMP*SDV(JV)*XMEAN(JV)
2764 U(CJJ,MP3)=TEMP
2774 U(CJJ,MP4)=TEMP
2784 1278 CONTINUE
2794 CC SET U(CJT,JS)=XX(JT,JV) AS IT SHOULD BE FOR JT=1,2,..,L9PLIT
2804 DO 1284 JT=1,N
2814 U(CJT,MP4)=XX(JT,JV)
2824 CONTINUE
2834 DO 1304 JT=1,L9PLIT
2844 DO 1298 JS=1,MP3
2854 U(CJT,JS)=U(CJT,MP4)
2864 CONTINUE
2874 1298 CONTINUE
2884 IF (LOGX.EQ.0) GO TO 1320
2894 DO 1314 JT=1,L9PLIT
2904 DO 1318 JS=1,MP4
2914 U(CJT,JS)=10.0**U(CJT,JS)
2924 CONTINUE
2934 1314 CONTINUE
2944 NTAPE=12+JV
2954 REWIND NTAPE
2964 DO 1334 K8=1,MP4
2974 WRITE (NTAPE,1430) (U(KT,K8),KT=1,L8P)
2984 CONTINUE
2994 END FILE NTAPE
3004 IF (NPRINT.EQ.0) GO TO 1350
3014 PRINT 1460
3024 DO 1344 K8=1,MP3
3034 PRINT 1460, (U(KT,K8),KT=1,L8P)
3044 CONTINUE
3054 1344 CONTINUE
3064 IF (PUNCH.EQ.0) GO TO 1378
3074 DO 1368 K8=1,MP3

```



```

66 FCST(1,I) = FCST(1,I)+TEMP
IF(LV2QP .GE. LV5) GO TO 88
KK = LV2QP+1
DO 75 I=KK,LV5
TEMP = 0.008
DO 78 J=1,IPD
L1 = I+J
IF(L1 .NE. 0) GO TO 65
TEMP = TEMP+DBLE(DARPB(J))
GO TO 78
76 CONTINUE
75 FCST(1,I) = FCST(1,I)+TEMP
88 TEMP = 0.008
K = K+1
I0 = LV(3)-1
I00 = LV(3)
I0P = I0+1
I022 = I0+100
IF(I00 .EQ. 0) GO TO 98
DO 85 I=I0P,I022
DO 85 PHAS(I) = ZERO
C
C
98 IF(LV1 .LT. LV2QP) GO TO 125
DO 120 I=LV2QP,LV1
TEMP1 = 0.008
TEMP2 = 0.008
IF(IPD .EQ. 0) GO TO 100
DO 95 J=1,IPD
TEMP1 = TEMP1+DBLE(DARPB(J))*DBLE(Z(I=J))
I02 = I022
DO 105 J=1,I0
TEMP2 = TEMP2+DBLE(PHAS(J))*DBLE(PHAS(J+I00))
IF(I0 .LT. 1) GO TO 115
DO 110 J=1,I0
I02 = I02+1
110 PHAS(I02+1) = PHAS(I02)
115 PHAS(I0P) = X(3)+PHAS(I0P)+TEMP1+TEMP2
120 TEMP = TEMP+DBLE(PHAS(I0P))*2
125 MNV = TEMP/LV1
TA = 1.0-ALPHA/2.0
CALL MDNRIS(TA,X,I,IPR)
IF(IPR .GT. 127) GO TO 195
S = X*SORT(HNV)
C
C
C COMPUTE THE CORRESPONDING DEVIATIONS
FROM EACH FORECAST FOR THE
PROBABILITY LIMITS
FCST(3,1) = 3
IF(LV5 .EQ. 1) GO TO 148
DO 135 I=2,LV5
TEMP = 0.008
L = I-1
DO 138 J=1,L
TEMP = TEMP+DBLE(FCST(1,J))*DBLE(FCST(1,J))
135 FCST(3,I) = S*DBORT(1.0DB0*TEMP)
148 K = KI
IF(K .EQ. 0) GO TO 155
DO 158 I=1,K
M1 = LV1+I
C
C
C COMPUTE THE ESTIMATE OF WHITE NOISE
VARIANCE
140 FCST(2,I) = FCST(2,I)+TEMP
145 FCST(2,I) = PHAS*TEMP
155 K = K+1
IF(K .GT. KI) GO TO 165
DO 160 I=K,KI
PHAC = FCST(2,I) + PHAC
165 IF(MOL .EQ. 0) GO TO 180
LVSP = LV3+1
DO 175 I=1,MOL
IF(I .GT. LV(3)) GO TO 188
TEMP = 0.008
DO 178 J=1,LV3
DO 178 J=1,LV3
178 TEMP = TEMP+DBLE(PHAS(J))*DBLE(PHAS(J+LV3P))
175 CONTINUE
180 IF(LV5 .EQ. 1) GO TO 9805
DO 190 I=2,LV5
KI = I-1
TEMP = 0.008
IF(IPD .EQ. 0) GO TO 9885
IF(KM1 .GT. IPD) KM1 = IPD
DO 185 J=1,KM1
IJ = J+J
TEMP = TEMP+DBLE(DARPB(J))*DBLE(FCST(2,IJ))
185 CONTINUE
190 CONTINUE
FCST(2,I) = FCST(2,I)+TEMP
GO TO 9885
195 IER = 129
GO TO 9888
200 IER = 130
GO TO 9888
205 IER = 131
9888 CONTINUE
9885 CALL UERTST(IER,SMFTCAST)
RETURN
END
FUNCTION FTFUNC (X,I,PAR)
C=FTFUNC-----8-----LIBRARY X-----
C
C FUNCTION = TO PROVIDE FUNCTIONAL COMMUNICATION BETWEEN
FTFUNC000 FTMAPS AND ZSYSTEM, NOT A STAND-ALONE ROUTINE
FTFUNC050 DUMMY = FTFUNC(X,I,PAR)
FTFUNC070 USAGE = DUMMY
FTFUNC080 PARAMETERS X(N) = AN INPUT VECTOR OF LENGTH N WHICH CONTAINS
C THE PRESENT APPROXIMATIONS TO AN
C UNNORMALIZED FORM OF THE MOVING AVERAGE
C PARAMETERS OF THE MODEL BEING ESTIMATED.
C N EQUALS THE NUMBER OF MOVING ESTIMATE.
C PARAMETERS PLUS 1.
C I = THE INDEX OF THE FUNCTION WHICH IS TO BE
C COMPUTED.
C PAR(N) = INPUT VECTOR OF LENGTH N (DESCRIBED ABOVE)
C CONTAINING COVARIANCES FROM IMSL ROUTINE
FTFUNC170

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FTR0560
FTR0590
FTR0600
FTR0610
FTR0620
FTR0630
FTR0640
FTR0650
FTR0660
FTR0670
FTR0680
FTR0690
FTR0700
FTR0710
FTR0720
FTR0730

      DO 30 I=M,LW
      Z(I-1) = Z(I)-Z(I-1)
      LM = LW - 15
      30 CONTINUE
      40 IF (IDI .EQ. 0) GO TO 9885
      DO 50 L=1,101
      DO 45 I=2,LW
      Z(I-1) = Z(I)-Z(I-1)
      LM = LW-1
      50 CONTINUE
      GO TO 9885
      9880 CONTINUE
      IER = 129
      CALL UERTST(IER,GMTRDIF)
      9885 RETURN
      END

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C=CGNOR-----8-----LIBRARY 3-----
C
C FUNCTION          = GENERATES PSEUDO NORMAL RANDOM NUMBERS.
C USAGE            = CALL CGNOR(SEED,N,R)
C PARAMETERS      = SEED = A DOUBLE PRECISION NUMBER, NOT OF THE FORM
C                   J/(2**K)(J,K INTEGERS) AND IN THE EXCLUSIVE
C                   RANGE (1.E-6,1.). SEED IS USED TO INITIATE
C                   THE GENERATION, AND ON EXIT, HAS BEEN
C                   REPLACED BY A NEW SEED FOR SUBSEQUENT USE.
C                   N = INPUT NUMBER OF DEVIATES TO BE GENERATED.
C                   R = OUTPUT VECTOR CONTAINING THE NORMAL
C                   PSEUDO RANDOM NUMBERS.
C PRECISION       = SINGLE
C RECD. INSL ROUTINES = CGUI,MERT1,UERTST
C LANGUAGE        = FORTRAN
C LATEST REVISION = FEBRUARY 28, 1973

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C=GGUI-----8-----LIBRARY 3-----
C
C FUNCTION          = GENERATION OF PSEUDO-RANDOM UNIFORM (0,1)
C                   DEVIATES (ONE MULTIPLIER)
C USAGE            = CALL GGUI(SEED,N,R)

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F7M1000
FTR0810
FTR0820
FTR0830
FTR0840
FTR0850
FTR0860
FTR0870
FTR0880
FTR0890
FTR0900
FTR0910
FTR0920
FTR0930
FTR0940
FTR0950
FTR0960
FTR0970
FTR0980
FTR0990
FTR1000

      SUBROUTINE FTRDIF (ID1,ID2,IP,IS,LZ,Z,SHIFT,LW,IER)
      C=PTRODIF-----8-----LIBRARY 3-----
      C
      C FUNCTION          = TIME SERIES TRANSFORMATION AND DIFFERENCING.
      C USAGE            = CALL FTRDIF(ID1,ID2,IP,IS,LZ,Z,SHIFT,LW,IER)
      C PARAMETERS      = ID1 = INPUT. ORDER OF NON-SEASONAL DIFFERENCE.
      C                   ID2 = INPUT. ORDER OF SEASONAL DIFFERENCE.
      C                   IP = INPUT. TRANSFORMATION EXPONENT.
      C                   IF IP = 0, A LOGRITHMIC TRANSFORMATION
      C                   IS PERFORMED ON THE INPUT SERIES.
      C                   IF IP NOT=0, AN EXPONENTIAL TRANSFORMATION
      C                   IS PERFORMED ON THE INPUT SERIES.
      C                   LZ = INPUT. LENGTH OF SEASONAL PERIOD.
      C                   Z(LZ) = INPUT AND OUTPUT VECTOR.
      C                   ON INPUT, Z CONTAINS THE TIME SERIES.
      C                   ON OUTPUT, Z CONTAINS THE TRANSFORMED AND
      C                   DIFFERENCED TIME SERIES.
      C                   SHIFT = OUTPUT. WHEN IP=0, SHIFT WILL BE
      C                   (THE NEGATIVE OF THE MINIMUM VALUE OF Z)
      C                   PLUS ONE. WHEN IP NOT=0, SHIFT WILL BE
      C                   ZERO.
      C                   LH = OUTPUT. TIME SERIES LENGTH. COMPUTED AS
      C                   LW = LZ-ID1-IS(ID2).
      C                   IER = ERROR PARAMETER.
      C                   N = 1 INDICATES ID1 AND/OR ID2 LESS THAN 0.
      C PRECISION       = SINGLE
      C RECD. INSL ROUTINES = UERTST
      C LANGUAGE        = FORTRAN
      C LATEST REVISION = MARCH 31, 1972

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C=CGUI-----8-----LIBRARY 3-----
C
C FUNCTION          = GENERATION OF PSEUDO-RANDOM UNIFORM (0,1)
C                   DEVIATES (ONE MULTIPLIER)
C USAGE            = CALL GGUI(SEED,N,R)

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C=GGUI-----8-----LIBRARY 3-----
C
C FUNCTION          = GENERATION OF PSEUDO-RANDOM UNIFORM (0,1)
C                   DEVIATES (ONE MULTIPLIER)
C USAGE            = CALL GGUI(SEED,N,R)

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C 1078 CONTINUE
IF (IPDQ(4),EQ,0) GO TO 1080
CALL STABPS (MK(N+1),DARPS,IPDQ(3),IPDQ(4),PMAS,MNV,MK(J),IER)
IF (IER,GT,0) GO TO 1250
IF (IPDQ(3),EQ,0) PMAC=MBAR
1080 CONTINUE
CALL CHECK (IPDQ(4),PMAS(1),IER)
IF (IER,GT,0) GO TO 1260
C 1081 CONTINUE
DO 1094 I=1,IR
MK(N+1)=0.0
1098 CONTINUE
IF (ID,GT,0) GO TO 1110
DO 1104 I=1,N
MK(I)=K(1)-MBAR
1108 CONTINUE
DO 1114 I=1,N
MPT=MAT
1118 CONTINUE
DO 1124 J=1,IP
MK(NPI)=MK(NPI)-DARPS(J)*K(I-J)
1128 CONTINUE
IF (IO,EQ,0) GO TO 1150
1138 CONTINUE
DO 1144 J=1,IO
MK(NPI)=MK(NPI)+PMAS(J)*K(NPI-J)
1148 CONTINUE
C 1150 CONTINUE
CALL FTAUTO (MK(N+1),N,120,10,15,MK(N+2),MK(N+22),MK(N+23),
1 MK(N+1),MK,MK)
C 1151 CONTINUE
T=0.0
DO 1164 I=1,120
T=T+K(N+1)+*2
1168 CONTINUE
X=N*N-IPDQ(5)
T=T*X
X=M*28-IP-IO
CALL MOCH (T,XM,CHI,IER)
IF ((1,9-CHI),GE,ALPHA(1)) GO TO 1170
GO TO 1270
C 1170 CONTINUE
ALPHA(1)=1.0-CHI
IF (ID,GT,0) PMAC=PMAC*(1.0-0.2*IPDQ(6)/IPDQ(2)-0.1)
IF (IPDQ(3),EQ,0) GO TO 1190
DO 1184 I=1,IP
MK(I)=DARPS(I)
1188 CONTINUE
1198 CONTINUE
IF (IPDQ(4),EQ,0) GO TO 1210
DO 1204 I=1,IO
MK(N+1)=PMAS(I)
1208 CONTINUE
1218 CONTINUE
CALL FTAUT (Z,MK,MK(N+1),PMAC,ALPHA(2),IPDQ(2),DARPS,FCST,MNV,
1 IERX)

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C
  IF (IPDG(7).LE.0) GO TO 1228
  SIMULATIONS
  M=IPDG(7)
  J=IP+ID
  K=MAX(7,1)
  DO 1228 I=1,M
  PMAC=TEMP+M*EAK(I)
  IF (ID.GT.0) PMAC=PMAC*(1.0-(0.2*IPDG(6))/IPDG(2)+0.1)
  CALL FTGENI (DARPS,PHAS,PMAC,Z(N=K),MNV,SEED,0,10,IPDG(6),8IN(1,
  1),MK)
  1220 CONTINUE
  RETURN
  1230 CONTINUE
  ALPHA(1)=1.0
  RETURN
  1240 CONTINUE
  ALPHA(1)=2.0
  RETURN
  1250 CONTINUE
  ALPHA(1)=3.0
  RETURN
  1260 CONTINUE
  ALPHA(1)=4.0
  RETURN
  1270 CONTINUE
  ALPHA(1)=5.0
  RETURN
  END

C SUBROUTINE IDENTN (Z,IND,RLIMIT,DISCON,ARF,IPARM,LOGX,NPRINT)
C FUNCTION
C Z IDENTIFY IP, ID AND IQ OF AN ARMA(IP,TD,IQ) MODEL
C IND(1)
C 1, LENGTH OF SERIES
C 2, LENGTH OF PROJECTION (MAXIMUM 600)
C 3, NUMBER OF SETS OF PROJECTION DESIRED (MAXIMUM 10)
C RLIMIT
C MAXIMUM ALLOWABLE RANGE OF PROJECTION, IT IS A FUNCTION OF
C DISCON
C DISCOUNT FACTOR (0.10 TO 0.50) IN COMPUTING DISCOUNTED
C ARF
C SIGNIFICANCE LEVEL (AFAL)
C -OUTPUT VECTOR GIVE FIRST FIVE HIGHEST DISCOUNTED
C SIGNIFICANCE LEVELS
C IPARM
C -OUTPUT MATRIX GIVES THE P, D, AND Q FOR THE TOP FIVE
C LOCK
C ARMA(P,D,Q) MODELS RANKED DECREASINGLY BY ARF VALUES
C LOGX
C 1, NATURAL LOG TRANSFORMATION OF Z IS DESIRED
C NPRINT
C 0, NO PRINT OUT IS DESIRED
C 1, PRINT OUT IS DESIRED
C
C=====
C DIMENSION Z(1),ALPHA(2),DARPS(5),PMAS(5),FCST(13,750),
C 1 SIM(750,5),MK(1500),IPDG(7)
C DIMENSION ZLAST(1,14),NAMES(14),NP(10),NG(14),ND(14)
C DIMENSION ALFA(14),ARF(5),RANGEC(14),RANGER(14),
C 1IND(3),IPARM(5,3)
C DOUBLE PRECISION SEED
C DATA NG/3,2,3,2,1,3,2,1,0,2,1,0,1,0/
C ND/1,1,1,2,3,0,1,2,3,0,1,2,0,1/
C NSEED(1)
C LPROJ=IND(2)
C SEARCH FOR OPTIMAL ID AS NDD
C CALL FINDD (Z,N,24,NDD,6,90)
C INITIALIZE PARAMETERS TO BE USED IN 6T8IMP
C

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  1S750
  IPDG(1)=1
  IPDG(2)=N
  IPDG(6)=LPROJ
  IPDG(7)=NSET
  DO 1080 I=1,14
  SEED=0.15678943250708
  ALPHA(1)=0.01
  ALPHA(2)=0.05
  IPDG(3)=N*IPDG(1)
  IPDG(4)=N*IPDG(1)
  IPDG(5)=NDD
  DO 1010 IN=1,5
  DARPS(IN)=0.0
  PHAS(IN)=0.0
  1010 CONTINUE
  PMAC=0.0
  FCST(2,LPROJ)=0.0
  ZLAST(1,1)=Z(N)
  IF (LOGX.NE.0) ZLAST(1,3)=EXP(Z(N))
  DO 1020 I=1,NSET
  ZLAST(1+I,1)=0.0
  1020 CONTINUE
  CALL 6T8IMP (Z,IPDG,SEED,ALPHA,DARPS,PHAS,PMAC,MNV,FCST,SIM,IS,MK,IDE 590
  1 IER)
  ALFA(1)=ALPHA(1)
  ND(I)=IPDG(5)
  PRINT 1130, I, ((IPDG(J),J=3,5),ALPHA(1),MNV,PHAS,CDARPS(J),J=,
  C 1 4),(PHAS(J),J=1,3)
  C 1 IF (IER.GT.0) ALFA(1)=LE.0) GO TO 1068
  CC COMPUTE THE RANGE OF PROJECT AS RANGEP
  XMIN=SIM(1,1)
  XMAX=SIM(1,1)
  RANGEP(1)=0
  DO 1040 J=1,NSET
  DO 1030 I=1,LPROJ
  IF (SIM(I,J).GT.XMAX) XMAX=SIM(I,J)
  IF (SIM(I,J).LT.XMIN) XMIN=SIM(I,J)
  RPXMAX=XMIN
  1030 CONTINUE
  IF (RP.GT.RANGEP(1)) RANGEP(1)=RP
  CC IF (RANGEP(1) LE. RLIMIT) GO TO 1048
  ALFA(1)=6.0
  GO TO 1068
  1048 CONTINUE
  IF (LOGX.EQ.0) GO TO 1068
  FCST(2,LPROJ)=EXP(FCST(2,LPROJ))
  DO 1050 J=1,NSET
  SIM(LPROJ,J)=EXP(SIM(LPROJ,J))
  1050 CONTINUE
  1068 CONTINUE
  ZLASTINSET(2,1)=FCST(2,LPROJ)
  ZLAST(I,1)=SIM(LPROJ,1)
  1070 CONTINUE
  1080 CONTINUE
  I=NBSE+1
  DO 1080 I=1,14
  C 1 PRINT 1140, K, NP(K), ND(K), ALFA(K), ZLAST(I,1), (ZLAST(J,K),
  C 1 J=2,7)
  C 1090 CONTINUE

```



```

PLO 940
PLO 950
PLO 960
PLO 970
PLO 980
PLO 990
PLO1000
PLO1010
PLO1020
PLO1030
PLO1040
PLO1050
PLO1060
PLO1070
PLO1080
PLO1090
PLO1100
PLO1110
PLO1120
PLO1130
PLO1140
PLO1150
PLO1160
PLO1170
PLO1180
PLO1190
PLO1200
PLO1210
PLO1220
PLO1230
PLO1240
PLO1250
PLO1260
PLO1270
PLO1280
PLO1290
PLO1300
PLO1310
PLO1320
PLO1330
PLO1340
PLO1350
PLO1360

```

```

LX=1
DO 1130 L=1,13
  ISAVE(LX)=IPER
  IPLOT(LX)=IPER
  LX=LX+10
1130 CONTINUE
KDEX=IBEGIN
DO 1160 I=1,N
  IF (K,GT,1)AND(K,LT,N) GO TO 1159
  DO 1140 L=1,12
    IPLOT(L)=IPER
  1140 CONTINUE
  1150 CONTINUE
  DO 1160 I=1,NV
    TEMP=(K,I)
    ITEMP=(TEMP-STRT)/STEP+.5
    LX=ITMP*ISTEP+1
    IF (LX,GT,121) LX=121
    IF (LX,LT,1) LX=1
    IPLOT(LX)=LTR(I)
  1160 CONTINUE
  WRITE (6,1260) KDEX,IPLOT
  KDEX=1+KDEX
  LAG=LAG+1
  DO 1170 L=1,121
    IPLOT(L)=ISAVE(L)
  1170 CONTINUE
  1180 CONTINUE
  WRITE (6,1270)
  RETURN
1170 CONTINUE
  WRITE (6,1280)
  RETURN
  WRITE (101,101,9A10)
1200 FORMAT (10X,20QUANTITIES SCALED BY ,F0.0)
1210 FORMAT (10)
1220 FORMAT (10)
1230 FORMAT (10)
1240 FORMAT (10)
1250 FORMAT (10)
1260 FORMAT (10)
1270 FORMAT (10)
1280 FORMAT (10)
END

```

```

LE1F0010
LE1F0020
LE1F0030
LE1F0040
LE1F0050
LE1F0060
LE1F0070
LE1F0080
LE1F0090
LE1F0100
LE1F0110
LE1F0120
LE1F0130
LE1F0140
LE1F0150
LE1F0160
LE1F0170
LE1F0180
LE1F0190
LE1F0200
LE1F0210
LE1F0220
LE1F0230
LE1F0240
LE1F0250
LE1F0260
LE1F0270
LE1F0280
LE1F0290
LE1F0300
LE1F0310
LE1F0320
LE1F0330
LE1F0340
LE1F0350
LE1F0360
LE1F0370
LE1F0380
LE1F0390
LE1F0400
LE1F0410
LE1F0420
LE1F0430
LE1F0440
LE1F0450
LE1F0460
LE1F0470
LE1F0480
LE1F0490
LE1F0500
LE1F0510
LE1F0520

```

```

SUBROUTINE LEQ1F (A,M,N,IA,B,IDGT,MKAREA,IER)
C-----LIBRARY 3-----
C-----8-----
C
C FUNCTION
C   = LINEAR EQUATION SOLUTION - FULL STORAGE
C   MODE = SPACE ECONOMIZER SOLUTION.
C   CALL LEQ1F (A,M,N,IA,B,IDGT,MKAREA,IER)
C   INPUT MATRIX OF DIMENSION N BY N CONTAINING
C   PARAMETERS A
C   THE COEFFICIENT MATRIX OF THE EQUATION
C   AX = B.
C   ON OUTPUT, A IS REPLACED BY THE LU
C   DECOMPOSITION OF A ROWWISE PERMUTATION OF
C   A.
C   M = NUMBER OF RIGHT-HAND SIDES.(INPUT)
C   N = ORDER OF A AND NUMBER OF ROWS IN B.(INPUT)
C   IA = NUMBER OF ROWS IN THE DIMENSION STATEMENT
C   FOR A AND B IN THE CALLING PROGRAM. (INPUT)
C   B = INPUT MATRIX OF DIMENSION N BY M CONTAINING
C   RIGHT-HAND SIDES OF THE EQUATION AX = B.
C   ON OUTPUT, THE N BY M SOLUTION X REPLACES B.
C   IDGT = INPUT OPTION.
C   IF IDGT IS GREATER THAN 0, THE ELEMENTS OF
C   A AND B ARE ASSUMED TO BE CORRECT TO IDGT
C   DECIMAL DIGITS AND THE ROUTINE PERFORMS
C   AN ACCURACY TEST.
C   IF IDGT EQUALS ZERO, THE ACCURACY TEST IS
C   BYPASSED.
C   MKAREA = WORK AREA OF DIMENSION GREATER THAN OR EQUAL
C   TO N.
C   IER = ERROR PARAMETER
C   TERMINAL ERROR = 120+*.
C   N = 1 INDICATES THAT A IS ALGORITHMICALLY
C   SINGULAR. (SEE THE CHAPTER L PRELUDE).
C   HARMING ERROR = 32+*.
C   N = 2 INDICATES THAT THE ACCURACY TEST
C   FAILED.
C   THE COMPUTED SOLUTION MAY BE IN ERROR
C   BY MORE THAN CAN BE ACCOUNTED FOR BY
C   THE UNCERTAINTY OF THE DATA.
C   THIS WARNING CAN BE PRODUCED ONLY IF
C   IDGT IS GREATER THAN 0 ON INPUT.
C   SEE CHAPTER L PRELUDE FOR FURTHER
C   DISCUSSION.
C
C PRECISION = SINGLE.
C READ. IM8L ROUTINES = LUDATF,LUELMF,UERT8T
C LANGUAGE = FORTRAN
C-----8-----
C LATEST REVISION = AUGUST 19, 1973
C
C DIMENSION A(IA,1),B(IA,1),MKAREA(1)
C INITIALIZE IER
C IER=0
C

```



```

LUDAI1780
LUDAI1710
LUDAI1720
LUDAI1730
LUDAI1740
LUDAI1750
LUDAI1760
LUDAI1770
LUDAI1780
LUDAI1790
LUDAI1800
LUDAI1810
LUDAI1820
LUDAI1830
LUDAI1840
LUDAI1850
LUDAI1860
LUDAI1870
LUDAI1880
LUDAI1890
LUDAI1900
LUDAI1910
LUDAI1920
LUDAI1930
LUDAI1940
LUDAI1950

          95      GO TO 90
          CONTINUE
          IF (JPI .GT. N) GO TO 185      DIVIDE BY PIVOT ELEMENT U(J,J)
          P = LU(J,J)
          DO 100 I=JPI,N
             LU(I,J) = LU(I,J)/P
          100 CONTINUE
          185 CONTINUE
          185 CONTINUE
          PERFORM ACCURACY TEST
          IF (IDGT .GE. 8) GO TO 9885
          P = 3*PI*J
          WA = P*W*KL
          IF (ABS(1.0-ABS(IDGT)) .NE. MA) GO TO 9885
          IER = 34
          GO TO 9888
          ALGORITHMIC SINGULARITY
          118 IER = 129
          D1 = ZERO
          D2 = ZERO
          9888 CONTINUE
          PRINT ERROR
          9885 CALL UERTST(IER,6,LU(1:J))
          RETURN
          END

SUBROUTINE LU(ELMP (A,B,IPVT,N,IA,X)
C=LUELMP=====LIBRARY 3=====
C
C FUNCTION = ELIMINATION PART OF SOLUTION OF AX=B
C USAGE = CALL LU(ELMP (A,B,IPVT,N,IA,X)
C PARAMETERS A = THE RESULT, LU, COMPUTED IN THE SUBROUTINE
C              *LU(1:J), WHERE L IS A LOWER TRIANGULAR
C              UPPER TRIANGULAR, L AND U ARE STORED AS A
C              SINGLE MATRIX, A, AND THE UNIT DIAGONAL OF
C              L IS NOT STORED
C              B = B IS A VECTOR OF LENGTH N ON THE RIGHT HAND
C              SIDE OF THE EQUATION AX=B
C              IPVT = THE PERMUTATION MATRIX RETURNED FROM THE
C              SUBROUTINE *LU(1:J), STORED AS AN M LENGTH
C              VECTOR
C              N = ORDER OF A AND NUMBER OF ROWS IN B
C              IA = NUMBER OF ROWS IN THE DIMENSION STATEMENT
C              FOR A IN THE CALLING PROGRAM.
C              X = THE RESULT X
C              *PRECISION
C              *LANGUAGE
C              *PORTRAN
C              *LATEST REVISION = AUGUST 15, 1975
C
C DIMENSION A(IA,1),B(1),IPVT(1),X(1)
C          DO 5 I=1,N
C              5 X(I) = B(I)
C          IN = B

```

```

LUDAI1800
LUDAI1810
LUDAI1820
LUDAI1830
LUDAI1840
LUDAI1850
LUDAI1860
LUDAI1870
LUDAI1880
LUDAI1890
LUDAI1900
LUDAI1910
LUDAI1920
LUDAI1930
LUDAI1940
LUDAI1950
LUDAI1960
LUDAI1970
LUDAI1980
LUDAI1990
LUDAI2000
LUDAI2010
LUDAI2020
LUDAI2030
LUDAI2040
LUDAI2050
LUDAI2060
LUDAI2070
LUDAI2080
LUDAI2090
LUDAI2100
LUDAI2110
LUDAI2120
LUDAI2130
LUDAI2140
LUDAI2150
LUDAI2160
LUDAI2170
LUDAI2180
LUDAI2190
LUDAI2200
LUDAI2210
LUDAI2220
LUDAI2230
LUDAI2240
LUDAI2250
LUDAI2260
LUDAI2270
LUDAI2280
LUDAI2290
LUDAI2300
LUDAI2310
LUDAI2320
LUDAI2330
LUDAI2340
LUDAI2350
LUDAI2360
LUDAI2370
LUDAI2380
LUDAI2390
LUDAI2400
LUDAI2410
LUDAI2420
LUDAI2430
LUDAI2440
LUDAI2450
LUDAI2460
LUDAI2470
LUDAI2480
LUDAI2490
LUDAI2500

          25      IF (AI .EQ. ZERO) AI = B(1A)
          TEST = MI/AI
          IF (TEST .GT. WREL) WREL = TEST
          GO TO 35
          30      IF (MI .LT. 1) GO TO 35      WITHOUT ACCURACY
          DO 30 K=1,MI
             SUM = SUM+LU(I,K)*LU(K,J)
          CONTINUE
          LU(I,J) = SUM
          P = ZERO
          40      GO TO I=J,N      COMPUTE U(J,J) AND L(I,J), I=J+1,...
          SUM = LU(I,J)
          IF (IDGT .GE. 8) GO TO 55
          WITH ACCURACY TEST
          AI = ABS(SUM)
          WI = ZERO
          IF (MI .LT. 1) GO TO 50
          DO 45 K=1,MI
             Y = LU(I,K)*LU(K,J)
             SUM = SUM+Y
             WI = WI+ABS(Y)
          CONTINUE
          LU(I,J) = SUM
          WI = WI*ABS(SUM)
          IF (AI .EQ. ZERO) AI = B(1A)
          TEST = WI/AI
          IF (TEST .GT. WREL) WREL = TEST
          GO TO 65
          WITHOUT ACCURACY TEST
          55      IF (MI .LT. 1) GO TO 65
          DO 60 K=1,MI
             SUM = SUM+LU(I,K)*LU(K,J)
          CONTINUE
          LU(I,J) = SUM
          60      Q = EQUIL(I)*ABS(SUM)
          IF (P .GE. 8) GO TO 70
          P = Q
          IMAX = I
          CONTINUE
          TEST FOR ALGORITHMIC SINGULARITY
          IF (RN*P .EQ. RN) GO TO 119
          IF (J .EQ. IMAX) GO TO 88
          INTERCHANGE ROWS J AND IMAX
          D1 = -D1
          DO 75 K=1,N
             P = LU(IMAX,K)
             LU(IMAX,K) = LU(J,K)
             LU(J,K) = P
          CONTINUE
          EQUIL(IMAX) = EQUIL(J)
          75      IPVT(J) = IMAX
          D1 = D1*LU(J,J)
          IF (ABS(D1) .LE. ONE) GO TO 88
          D1 = D1*SIXTH
          D2 = D2*FOUR
          GO TO 85
          80      IF (ABS(D1) .GE. SIXTH) GO TO 95
          D1 = D1*SIXTH
          D2 = D2*FOUR

```

```

DO 26 I=1,N
  IP = IPVT(I)
  SUM = X(IP)
  X(IP) = X(I)
  IF (IM .EQ. 0) GO TO 15
  IM = I-1
  DO 10 J=1,IM
    SUM = SUM+X(I,J)*X(J)
  10 CONTINUE
  GO TO 20
  IF (SUM .NE. 0.) IM = I
  20 X(I) = SUM
  C
  DO 30 I=1,N
    I = N+1-IB
    IP = I+1
    SUM = X(I)
    IF (IP1 .GT. N) GO TO 30
    DO 25 J=IP1,N
      SUM = SUM+X(I,J)*X(J)
    25 CONTINUE
  30 X(I) = SUM/A(I,I)
  RETURN
  END

```

```

LUFF0330
LUFF0340
LUFF0350
LUFF0360
LUFF0370
LUFF0380
LUFF0390
LUFF0400
LUFF0410
LUFF0420
LUFF0430
LUFF0440
LUFF0450
LUFF0460
LUFF0470
LUFF0480
LUFF0490
LUFF0500
LUFF0510
LUFF0520
LUFF0530
LUFF0540
LUFF0550
LUFF0560

```

```

DO 26 I=1,N
  IP = IPVT(I)
  SUM = X(IP)
  X(IP) = X(I)
  IF (IM .EQ. 0) GO TO 15
  IM = I-1
  DO 10 J=1,IM
    SUM = SUM+X(I,J)*X(J)
  10 CONTINUE
  GO TO 20
  IF (SUM .NE. 0.) IM = I
  20 X(I) = SUM
  C
  DO 30 I=1,N
    I = N+1-IB
    IP = I+1
    SUM = X(I)
    IF (IP1 .GT. N) GO TO 30
    DO 25 J=IP1,N
      SUM = SUM+X(I,J)*X(J)
    25 CONTINUE
  30 X(I) = SUM/A(I,I)
  RETURN
  END

```

```

SUBROUTINE MDCM (CS,DF,P,IER)
  C=MDCM-----LIBRARY 3-----
  C FUNCTION
  C USABE MDCM (CS,DF,P,IER)
  C PARAMETERS CS
  C DF
  C P
  C IER
  C PRECISION
  C READ,INBL ROUTINES = UERTL,WERF,MGAMMA
  C LANGUAGE = FORTRAN
  C LATEST REVISION = APRIL 3,1974
  C DATA
  C PUNCM(A,Z) = EXP(A*ALOG(Z)-Z)
  C IF (DF .GE. .5 .AND. DF .LE. 2.E5 .AND. CS .GE. 8.E6) GO TO 5
  IER=1
  PARINP

```

```

SUBROUTINE MDCM (CS,DF,P,IER)
  C=MDCM-----LIBRARY 3-----
  C FUNCTION
  C USABE MDCM (CS,DF,P,IER)
  C PARAMETERS CS
  C DF
  C P
  C IER
  C PRECISION
  C READ,INBL ROUTINES = UERTL,WERF,MGAMMA
  C LANGUAGE = FORTRAN
  C LATEST REVISION = APRIL 3,1974
  C DATA
  C PUNCM(A,Z) = EXP(A*ALOG(Z)-Z)
  C IF (DF .GE. .5 .AND. DF .LE. 2.E5 .AND. CS .GE. 8.E6) GO TO 5
  IER=1
  PARINP

```


HDCI0610
HDCI0620
HDCI0630
HDCI0640
HDCI0650
HDCI0660
HDCI0670
HDCI0680
HDCI0690
HDCI0700
HDCI0710
HDCI0720
HDCI0730
HDCI0740
HDCI0750
HDCI0760
HDCI0770
HDCI0780
HDCI0790
HDCI0800
HDCI0810
HDCI0820
HDCI0830
HDCI0840
HDCI0850
HDCI0860
HDCI0870
HDCI0880
HDCI0890
HDCI0900
HDCI0910
HDCI0920
HDCI0930
HDCI0940
HDCI0950
HDCI0960
HDCI0970
HDCI0980
HDCI0990
HDCI1000
HDCI1010
HDCI1020
HDCI1030
HDCI1040
HDCI1050
HDCI1060
HDCI1070
HDCI1080
HDCI1090
HDCI1100
HDCI1110
HDCI1120
HDCI1130
HDCI1140
HDCI1150
HDCI1160
HDCI1170
HDCI1180
HDCI1190
HDCI1200

```

NCNT = NCNT + 1
IF (NCNT .GT. NCT) GO TO 98
IF (P1 .EQ. 0) 25,985,130
25 X = X + DX
IF (X .GT. 1)
  IF (IST .EQ. 0) GO TO 5
  GO TO 35
30 X = X - DX
IF (X .GT. 1)
  IF (IEM .EQ. 0) GO TO 5
  XR = CSR
  XL = X
  GO TO 40
35 XL = CSR
  XR = X
  C 40 EPSP = 10.**(NSIG)
  IF (XL .GT. 0) XL = 0.
  CALL MOCH (XL,DF,P1,IER)
  IF (IER .NE. 0) GO TO 65
  FXL = P1 * P
  CALL MOCH (FXL,DF,P1,IER)
  IF (IER .NE. 0) GO TO 65
  FXR = P1 * P
  IF (FXL .GT. 0) FXR = 0.
  IF (FXL .GT. 0) X = XL
  IF (FXR .GT. 0) X = XR
  GO TO 985
45 IF (DF .LE. 2. * OR. P .GT. .98) GO TO 50
  REGULA = 0
  CALL REGULA (X,DF,P1,IER)
  IF (IER .NE. 0) GO TO 65
  X = XL + FXL * (XR - XL) / (FXL - FXR)
  GO TO 55
  C 50 SECTION METHOD
  50 X = (XL + XR) / 2
  55 CALL MOCH (X,DF,P1,IER)
  IF (IER .NE. 0) GO TO 65
  FCS = P1 * P
  IF (ABS(FCS) .GT. EPS) GO TO 68
  GO TO 985
68 IF (FCS .GT. 0)
  XR = X
  FXR = FCS
  GO TO 78
  IF (FCS .LT. 0)
  FXL = X
  FCS = FCS
  GO TO 78
78 IF (XR - XL) .GT. EPS * ABS(XR) GO TO 75
75 IC = IC + 1
  IF (IC .LE. ITHAX) GO TO 45
  IER = 132
  C 80 IER = 130
  IER = 130
  GO TO 980B
  C 85 IER = 131
  IER = 131
  GO TO 980B
  980B CONTINUE
  980B CALL UERTBT (IER,AMDCHE)
  980B RETURN
  END
  
```

```

HDCI0610
HDCI0620
HDCI0630
HDCI0640
HDCI0650
HDCI0660
HDCI0670
HDCI0680
HDCI0690
HDCI0700
HDCI0710
HDCI0720
HDCI0730
HDCI0740
HDCI0750
HDCI0760
HDCI0770
HDCI0780
HDCI0790
HDCI0800
HDCI0810
HDCI0820
HDCI0830
HDCI0840
HDCI0850
HDCI0860
HDCI0870
HDCI0880
HDCI0890
HDCI0900
HDCI0910
HDCI0920
HDCI0930
HDCI0940
HDCI0950
HDCI0960
HDCI0970
HDCI0980
HDCI0990
HDCI1000
HDCI1010
HDCI1020
HDCI1030
HDCI1040
HDCI1050
HDCI1060
HDCI1070
HDCI1080
HDCI1090
HDCI1100
HDCI1110
HDCI1120
HDCI1130
HDCI1140
HDCI1150
HDCI1160
HDCI1170
HDCI1180
HDCI1190
HDCI1200
SUBROUTINE MOCH (P,DF,X,IER)
  C MOCH: INVERSE CHI-SQUARED PROBABILITY DISTRIBUTION
  C MOCH: FUNCTION (P,DF,X,IER)
  C MOCH: CALL MOCH (P,DF,X,IER)
  C MOCH: INPUT PROBABILITY IN THE EXCLUSIVE RANGE
  C MOCH: (0,1)
  C MOCH: INPUT VALUE CONTAINING NUMBER OF DEGREES OF
  C MOCH: FREEDOM. DF MUST BE IN THE EXCLUSIVE RANGE
  C MOCH: (.5,20000)
  C MOCH: OUTPUT CHI-SQUARED VALUE, SUCH THAT A RANDOM
  C MOCH: VARIABLE, DISTRIBUTED AS CHI-SQUARED WITH
  C MOCH: DF DEGREES OF FREEDOM, WILL BE LESS THAN OR
  C MOCH: EQUAL TO X WITH PROBABILITY P.
  C MOCH: IER = 0: SUCCESS
  C MOCH: IER = 1: TERMINAL ERROR
  C MOCH: IER = 2: N = 1 MEANS THAT BOUNDS WHICH ENCLOSED P
  C MOCH: COULD NOT BE FOUND WITHIN 26 (NCT)
  C MOCH: IER = 3: N = 2 MEANS THAT AN ERROR OCCURRED IN
  C MOCH: ITERATIONS
  C MOCH: IER = 4: N = 3 MEANS THAT AN ERROR OCCURRED IN MOCH
  C MOCH: FOUND WITHIN 50 (ITHAX) ITERATIONS. 80
  C MOCH: THAT THE ABSOLUTE VALUE OF P1P HAS NOT
  C MOCH: .LE. EPS. (P1P CALCULATED PROBABILITY AT X,
  C MOCH: EPS = .0001)
  C MOCH: IER = 5: SINGLE
  C MOCH: REGO, IML ROUTINES = MOCH, MERFL, UERTBT, MERF, MGAMMA
  C MOCH: LANGUAGE = FORTRAN
  C MOCH: LATEST REVISION = SEPTEMBER 1, 1973
  C MOCH: EPS/.0001, ITHAX/50, NCT/20, NSIG/5/
  C MOCH: ESTIMATE STARTING X
  C MOCH: CALL MOCH (P,DF,X,IER)
  C MOCH: IF (IER .NE. 0) GO TO 68
  C MOCH: D = SORT(DF)
  C MOCH: X = DF*(1. - DFF + XP/2)**3
  C MOCH: IF (DF .GT. 48) GO TO 985
  C MOCH: FIND BOUNDS (IN X) WHICH ENCLOSE P
  C MOCH: NCNT = 0
  C MOCH: IST = 0
  C MOCH: ISM = 0
  C MOCH: DX = X * .125
  C MOCH: 5 IP (X) 18,15,28
  C MOCH: 18 X = 9.8
  C MOCH: DX = -DX
  C MOCH: GO TO 28
  C MOCH: 15 DX = .1
  C MOCH: 28 CALL MOCH (X,DF,P1,IER)
  C MOCH: DX = DX + DX
  C MOCH: IF (IER .NE. 0) GO TO 85
  C MOCH: CSR = X
  
```



```

35 IF(IQUIT .EQ. N) GO TO 148
36 I = K
37 IP=IPART+I
38 ITEMP = WALKSUB+I) + PHI
39 HOLD = X(ITEMP)
40 ET = .001*ABS(HOLD)
41 IF (ABS(HOLD) .LT. PREC) ET=ETADELTA
42 H = AMIN((PHAX,ETA)
43 IF (H .LT. PREC) H=PREC
44 X(ITEMP)=HOLD+H
45 IF (K .LE. 1) GO TO 45
46 KK = 2
47 GO TO 15
48 PPLUS=FX,K,PAR)
49 TOP=PLUS-E
50 WAIPI)=TOP/H
51 X(ITEMP)=HOLD
52 IF (I .LE. N) GO TO 48
53 IF (K .LT. N) GO TO 68
54 IP=IPART+N
55 IF (ABS(WAIPI)) .EQ. ZERO) GO TO 88
56 X(ITEMP) = -E/WAIPI) + X(ITEMP)
57 GO TO 168
58
59
60 KLKLSUR+K
61 LOOK= WAK(L) + PHI
62 KMAXLOOK
63 IP=IPART+K
64 DERMAX=ABS(WAIPI)
65 KPLUS = K+1
66 DO 65 I = KPLUS,N
67 TEST=ABS(WAIPIPART+I))
68 IF (TEST .LE. DERMAX) GO TO 65
69 DERMAX = TEST
70 KMAX=I
71 CONTINUE
72 IF (LOOK .EQ. KMAX) GO TO 75
73 LKMAX=LKSUB+KMAX
74 WALK(LKMAX)
75 WALK(KMAX)LOOK
76 IP=IPART+KMAX
77 XTEMP= WAIPI)
78 IPK=IPART+K
79 WAIPI)=WAIPIK)
80 WAIPIK)=XTEMP
81 IF (K .LT. 2) GO TO 75
82 KMIN=K-1
83 I1 = 0
84 DO 78 I=1,KMIN
85 L= ((I1)+(N2 -I))/2=1
86 JEL+KMAX
87 XTEMP= WAIPI)
88 J=JL+K
89 WAIPI)=WAIPIJ)
90 WAIPIJ)=XTEMP
91 I1 = I
92 CONTINUE
93 IF ( ABS(WAIPIPART+K)) .NE. ZERO) GO TO 88
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APPENDIX B
DESCRIPTION OF COMPUTER PROGRAM MVPROJ

INTRODUCTION

1. Origin of Program. This program was developed by Shin Chang of The Center for Research in Water Resources, The University of Texas at Austin under the supervision of Leo R. Beard, Technical Director. Subroutines obtained from other sources are BMDOIM of Biomedical Computer Programs (BMD) and all those subroutines used in Program SVPROJ.

2. Capability of the MVPROJ Program. MVPROJ is capable of performing the following jobs:
 - a. Construction of the principal components of a set of original variables.
 - b. For each of these principal components:
 1. Identification of p , d , and q of the best-fit ARIMA(p,d,q) model for each principal component being analyzed.
 2. Estimation of parameters for the models identified in 1.
 3. Generation of 51 independent sets of projections for each principal component.
 - c. Computation of 51 projections for each of the original variables as functions of projections of all principal components.
 - d. Plotting up to 7 selected projections of each of the original variables in one chart for each variable.

3. Hardware and Software Requirements. This program has been developed and tested on the CDC 64/6600 computer system. Thirteen disk files are required. The Fortran IV compiler and the IMSL library are used.

DESCRIPTION OF THE PROGRAM

4. Program Organization. The MVPROJ program consists of a main program and 25 subroutines. All input data and input control parameters are read

in the main program from punched cards. The program will make a normal exit (unless errors occur) only in the main program.

5. Method of Computation. The computation procedures and equations used in this program are described in Chapter 5 of the main report under the title "Multivariate Projection Model."

6. Description of Subroutines and Variables. The function of each subroutine and the definitions of variables in each subroutine are given in the program listings. However, subroutines used in SVPROJ are not repeated here.

7. Dimensional Limitations of the Program. For practical purpose, the following limitations are set in the program:

<u>Variable</u>	<u>Size</u>	<u>Definition</u>
NV	12	number of variables
LSPLIT	150	length of calibration variables
LPROJ	100	length of projection
NSET	51	sets of independent projections

These dimensions can be easily extended by increasing the dimension specification in related arrays, however, ultimate limitations are determined by the available control memory of the computer system used.

8. Control Parameters for Optional Jobs. The following parameters are used to control the execution of certain optional jobs in the program.

- ISPLIT = controls the execution of the split-record test
- LOGX = Controls the execution of the logarithmic (base 10) transformation of the original variables
- NPRINT = Controls the execution of the printed output of the selected projections (7 sets)
- NPUNCH = Controls the execution of the punched-card output of the selected projections (7 sets)
- NPLOT = Controls the execution of the plotted output of the selected projections (7 sets or 5 sets)

9. Preparation of Input Variables.

<u>Variable</u>		<u>Description</u>
IFMT	character	variable format for input data x
ISPLIT	1	split-record test is desired,
	0	not desired
IYRB	integer	beginning year of record of the calibration series
IYRE	integer	ending year of record of the calibration series
IYRP	integer	ending year of projection
LOGX	1	logarithmic (base 10) transformation of the variables is desired
	2	not desired
NPLOT	0	no plot is desired as output
	1	plots of data and 7 projections depicted in one chart are desired
	2	plots of data and 7 projections, and data and 5 projections, depicted in 2 charts are desired
	3	plots of data and 5 projections depicted in one chart are desired
	integer	number of variables
NV	integer	number of variables
WT (12)	real	relative weights to be applied to the last member of projection of each variable for ranking the 51 sets of projections.
	real	relative weights to be applied to the last member of projection of each variable for ranking the 51 sets of projections.
YLABEL (8)	character	name of variable to be appeared in the chart
X (150, 12)	real	input data of calibration series, the first dimension refers to time and the second dimension refers to variables, (X(j,i) = jth record of ith variable)
	real	input data of calibration series, the first dimension refers to time and the second dimension refers to variables, (X(j,i) = jth record of ith variable)

10. Preparation of Input Cards. The following types of input cards are required.

<u>Card</u>	<u>Format</u>	<u>Variable(s)</u>
A	16I5	NV, IYRB, IYRE, IYRP, LOGX, ISPLIT, NPRINT, NPUNCH, NPLOT

B	12F5.0	WT
C	8A10	IFMT
D	IFMT	X
E	—	job card for BMD01M
F	—	label card for BMD01M
G	IFMT	data format card for BMD01M
H	8A10	YLABEL

Preparation of E and F cards of BMD01M:

E card (Job card):

col. 1-6	Problem (mandatory)
col. 7-12	Alphanumeric job code
col. 13,14	Number of original variables ($2 \leq NV \leq 12$)
col. 15-17	Length of calibration variables ($3 \leq LSPLIT \leq 150$)
col. 18-20	Blank
col. 21-23	Yes
col. 24-28	Blank
col. 29-30	Number of variables to be labeled
col. 31-68	Blank
col. 68-70	07 (data input is from logical BCD Tape 7)
col. 71,72	K Number of Variable Format Cards ($1 \leq K \leq 10$)

F Card (Labels card):

col. 1-6	LABELS (mandatory)
col. 7-10	The number of the variable to be named. The number must be right-justified
col. 11-16	The corresponding alphanumeric name
col. 17-20	The number of another variable
col. 21-26	The corresponding alphanumeric name.
⋮	

col. 67-70 The number of another variable
 col. 71-76 The corresponding alphanumeric name
 (up to 7 per card)

Example = The following Labels Card (F Card) is prepared for
 the case of Labelling X(j,2) as name 01 and X(j,5)
 as name 02.

Labels0002NAME010005NAME02.

Note 1. The number of a variable is i if the variable to be labeled is
 the ith variable of the input vector X(j,i).

2. More than one labels card can be used.

3. The number of labels appearing on all the labels cards must
 equal to the number of labels specified on the job card.

11. Arrangement of Input-deck. The following sequence of the input deck
 is used:

A B C	<u>D D D ...D</u>	E F G	<u>H H H ...H</u>
	X		YLABEL

12. Overview of Output. The following printout will be shown:

- a. Normalized and transformed (LOGX = 1) input data
- b. Correlation coefficient matrix
- c. Eigenvalues
- d. Eigenvectors
- e. For each principal component, a summary table showing the best-fit ARIMA (p,d,q) model, estimations of parameters, level of significance.

Optional printed and/or punched output of the 7 sets of projections and chart (s) of projections will be shown upon the specification of related output control parameters.

13. Program Listing. Source program listing for MVPROJ and its subroutines are shown in the following pages. Note that subroutines shared by both programs SVPROJ and MVPROJ will be shown in appendix A only.

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PROGRAM MVRPROJ (CALL TAPE1=1001, INPUT=1001, OUTPUT=1001, PUNCH=1001,
1 TAPE1=INPUT, PUNCH=OUTPUT, TAPE2=1001, TAPE3=1001, TAPE4=1001, TAPE5=
2 TAPE1, TAPE6=1001, TAPE7=1001, TAPE8=1001, TAPE9=1001, TAPE10=1001, TAPE11=
3 TAPE1, TAPE12=1001, TAPE13=1001, TAPE14=1001, TAPE15=1001, TAPE16=1001, TAPE17=1001, TAPE18=1001, TAPE19=1001, TAPE20=1001, TAPE21=1001, TAPE22=1001, TAPE23=1001, TAPE24=1001, TAPE25=1001, TAPE26=1001, TAPE27=1001, TAPE28=1001, TAPE29=1001, TAPE30=1001, TAPE31=1001, TAPE32=1001, TAPE33=1001, TAPE34=1001, TAPE35=1001, TAPE36=1001, TAPE37=1001, TAPE38=1001, TAPE39=1001, TAPE40=1001, TAPE41=1001, TAPE42=1001, TAPE43=1001, TAPE44=1001, TAPE45=1001, TAPE46=1001, TAPE47=1001, TAPE48=1001, TAPE49=1001, TAPE50=1001, TAPE51=1001, TAPE52=1001, TAPE53=1001, TAPE54=1001, TAPE55=1001, TAPE56=1001, TAPE57=1001, TAPE58=1001, TAPE59=1001, TAPE60=1001, TAPE61=1001, TAPE62=1001, TAPE63=1001, TAPE64=1001, TAPE65=1001, TAPE66=1001, TAPE67=1001, TAPE68=1001, TAPE69=1001, TAPE70=1001, TAPE71=1001, TAPE72=1001, TAPE73=1001, TAPE74=1001, TAPE75=1001, TAPE76=1001, TAPE77=1001, TAPE78=1001, TAPE79=1001, TAPE80=1001, TAPE81=1001, TAPE82=1001, TAPE83=1001, TAPE84=1001, TAPE85=1001, TAPE86=1001, TAPE87=1001, TAPE88=1001, TAPE89=1001, TAPE90=1001, TAPE91=1001, TAPE92=1001, TAPE93=1001, TAPE94=1001, TAPE95=1001, TAPE96=1001, TAPE97=1001, TAPE98=1001, TAPE99=1001, TAPE100=1001)
COMMON /JG/ISPLIT,NPRINT,NPLOT,NPUNCH,NSET,IYRB,IYRE,IYRQ,
COMMON /MODEL/MD,ND,NDP,ISPLIT,LPROJ,LEP
DOUBLE PRECISION RECD
COMMON /P/PC,PC2,PC3,PC4,PC5,PC6
DATA NDE(2)N/3,3,3,5,0
FOR A BETTER UNDERSTANDING OF THE ALGORITHM, THE FOLLOWING
C INTEGERS ARE USED.
C JI- THE J-TH TIME
C JI- THE J-TH VARIABLE
C JPC- THE J-TH PRINCIPLE COMPONENT
C JS- THE J-TH SIMULATION OF A VARIABLE OR A P=C
C D(J),J,PC=J-TH PRINCIPLE COMPONENT AT TIME J, AND
C JS=ITOT, J-TH SIMULATIONS OF THIS P=C
C JS=J, ORIGINATOR OF THIS P=C
C U(J,I,JS)=J-TH SIMULATION OF A VARIABLE AT TIME I
C NDE(J),
C READ(1200, NV, IYRB, IYRE, IYRQ, ISPLIT, NPUNCH, NPLOT,
C PRINT(1020, NV, IYRB, IYRE, IYRQ, ISPLIT, NPUNCH, NPLOT,
C NDE(J), JS)
C READ(1250, NP, (J), JI, I2)
C READ(1470, IFTM)
C DO 1010 JI=1, NV
C READ IFMT, X(J,I), I31, NV)
1010 CONTINUE, X(J,I), I31, I2)
C IF (LOCK.EQ.0) GO TO 1030
C DO 1020 JI=1, NV
C X(J,I)=ALOG10(X(J,I))
1020 CONTINUE
1030 CONTINUE
C STORE X(J,I) IN TAPE7 FOR BMD USESAS INPUT
C DO 1040 JI=1, NV
C WRITE (7,1040) (X(J,I), I31, NV)
1040 CONTINUE
C END FILE 7
C REWIND 7
C I TAPE=4
C CALL BMDRM
C TEMP=SQRT(FLOAT(N-1))
C DO 1050 I=1, NV
C SDV(I)=SDV(I)/TEMP
1050 CONTINUE
C IF LOG10-TRANSFORMATION IS USED IN THE PRINCIPAL COMPONENT
C ANALYSIS, THEN LOGX IS NOT EQUAL 0 (BY READ 2)
C C C FORECASTS AND SIMULATIONS OF THE P=C
C C C
C MSET=5
C MPM=SET+1
C MP2=SET+2
C MP3=SET+3
C MP4=SET+4
C LSPPLIT=N
C LPROJ=IYRQ-IYRE
C LSPPLIT=N/2+1
C LPROJ=N-LSPPLIT
C LSP=ISPLIT+LPROJ
C SEED=0.5234570548

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1844 J=INDEX(L)
1845 DO 1194 I=1,LPROJ
1846 IF (ISPY=0) GT,ZMAX) ZMAX=SIM(I,I,J)
1847 IF (SIM(I,I,J).LT.ZMIN) ZMIN=SIM(I,I,J)
1848 CONTINUE
1849 DO 1198 I=1,MP3
1850 IF (ISPY=0) EXP(G(J,I))+ZLO
1851 CONTINUE
1852 IF (ISPY=0) GO TO 1288
1853 IF (ISPY=0) GO TO 1288
1854 ALFA(I)=0.0
1855 CONTINUE
1856 C ITERATED CYCLE IN CASE THE FIRST CHOICE FROM IDENTN CAN
1857 C NOT GIVE SATISFIED SOLUTION
1858 C
1859 C IF (IER.GT.8.OR.ALFA(I).LE.0.0) GO TO 1218
1860 GO TO 1224
1861 C SOLUTION IS UNSATISFIED, ITERATED IF ISPY NOT EQUAL TO
1862 C ZERO AND ITERATION NUMBER JTT IS NOT EXCEED 5
1863 CONTINUE
1864 IF (ISPY.EQ.0) GO TO 1360
1865 IF (JTT.GE.5) GO TO 1360
1866 JTT=JTT+1
1867 GO TO 1120
1868 C
1869 C CONSTRUCTION OF G-ARRAY FOR PLOTTING
1870 GO TO 1120
1871 DO 1230 I=1,MP3
1872 DO 1230 J=1,LSPPLIT
1873 G(J,I)=G(J,MP3)
1874 CONTINUE
1875 DO 1250 J=1,LPROJ
1876 JJ=J+LSPPLIT
1877 L=NODE(I)
1878 I=INDEX(L)
1879 G(J,I)=SIM(J,I)
1880 CONTINUE
1881 TEMP=AG(JJ,2)
1882 TEMP=AG(JJ,MP1)
1883 G(JJ,MP2)=TEMP
1884 G(JJ,MP3)=FCST(2,J)
1885 CONTINUE
1886 REVERSE TRANSFORMATION OF VARIABLES IF APPLICABLE
1887 J=LSPPLIT+1
1888 IF (LOCK.EQ.0) GO TO 1280
1889 TEMP=ZUP-ZLO
1890 DO 1260 J=1,LSP
1891 DO 1260 I=1,MP3
1892 G(J,I)=EXP(G(J,I))+ZLO
1893 CONTINUE
1894 IF (TBOND.LE.0) GO TO 1288
1895 DO 1270 J=1,LSP
1896 DO 1270 I=1,MP3
1897 G(J,I)=TEMP/EXP(G(J,I)-ZLO)+ZLO
1898 CONTINUE
1899 IF ISPLIT=0, SET G(J,MP3)=G(J,MP3) FOR J GREATER THAN N
1900 IF (ISPLIT.NE.0) GO TO 1360
1901 DO 1290 J=1,LSP
1902 G(J,MP3)=G(J,MP3)
1903 CONTINUE
1904 1290 CONTINUE
1905 1296 CONTINUE

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1298 IPDG(7)=NSET
1299 JTT=1
1300 IF (ISPY=0) 1100,1100,1110
1301 CONTINUE
1302 IPDG(3)=IP
1303 IPDG(4)=IO
1304 IPDG(5)=ID
1305 GO TO 1130
1306 CONTINUE
1307 PARAMETERS IP,IO AND ID OF THE ARIMA (IP,IO,IO) MODEL ARE
1308 IDENTIFIED BY IDENTN ROUTINE USING WHOLE RECORD OF SERIE
1309 C
1310 INO(1)=N
1311 INO(2)=S*N
1312 INO(3)=S
1313 CALL IDENTN (Z,IND,RL,0,30,ARF,IPARM,1,1)
1314 CONTINUE
1315 IPDG(3)=IPARM(JTT,1)
1316 IPDG(4)=IPARM(JTT,3)
1317 IF (IPDG(1).GT.0) IPDG(5)=IPARM(JTT,2)
1318 CONTINUE
1319 SEED=0.52366705408
1320 ALPHA(1)=0.01
1321 ALPHA(2)=0.05
1322 IS=100
1323 DO 1150 IN=1,5
1324 DARPS(IN)=0.0
1325 PHAS(IN)=0.0
1326 CONTINUE
1327 PHAC=0.00
1328 CALL GTIMP (Z,IPDG,SEED,ALPHA,DARPS,PHAS,PHAC,MNV,FCST,SIM,IS,WK,
1329 I,IER)
1330 ALFA(I)=ALPHA(1)
1331 NP(I)=IPDG(3)
1332 NQ(I)=IPDG(4)
1333 MD(I)=IPDG(5)
1334 C RANK THE LAST MEMBERS OF SIMULATIONS SIM(I,J) INTO ZLAST
1335 FROM MAX TO MIN
1336 DO 1150 J=1,NSET
1337 WK(J)=SIM(LPROJ,J)
1338 CONTINUE
1339 SEARCH FOR THE MINIMUM AND THEN SET AMIN FOR RANKING OPERAT
1340 AMWK(1)
1341 DO 1160 J=1,NSET
1342 IF (WK(J).LT.A) AMWK(J)
1343 CONTINUE
1344 AMIN=AMWK(1)
1345 DO 1180 L=1,NSET
1346 AMIN=AMIN
1347 DO 1170 J=1,NSET
1348 IF (WK(J).LE.A) GO TO 1170
1349 AMWK(L)
1350 L=L+1
1351 CONTINUE
1352 WK(I)=AMIN
1353 INDEX(L)=I
1354 CONTINUE
1355 COMPUTE THE RANGE OF PROJECT AS RANGE
1356 RANGE(I)=0.0
1357 ZMAX=SIM(I,J)
1358 ZMIN=SIM(I,J)
1359 DO 1200 I=1,5
1360 L=NODE(I)

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DO 1818 I=1,N
TP=TPX(I)
ONE=1/TP
TMP=TMP+X(I)*ONE
1818 CONTINUE
IF (TP.GT.1. OR TMP.GT.1.) GO TO 1820
IF (NLT.2) GO TO 1830
TP=X(1)+X(13)-X(1)-X(2)
1820 CONTINUE
GO TO 1838
1838 IER=0
CALL PLOT (U,LSP,9,0,IYR8,YL8)
RETURN
END
SUBROUTINE FINDD (U,N,M,NDD,EPSON)
C-----
C FUNCTION -SEARCH THE ORDER OF DIFFERENCING D IN AN ARIMA.P
C D(0) MODEL SO THAT THE DIFFERENCED SERIES BECOMES
C U STATIONARY
C U -LENGTH OF VECTOR U
C M -LENGTH OF VECTOR U
C N -NO. OF AUTOCORRELATION COEFFICIENTS USED IN THE
C NDD -CHI-SQUARE STATISTIC
C EPSON -OUTPUT OPTIMAL ORDER OF DIFFERENCING D
C -CRITERION USED IN SELECTING NDD
C-----
DIMENSION U(1),CHI(3),AC(36),ACV(36),PACV(36),WKAREA(120),X(120)
DO 1818 J=1,N
X(J)=U(J)
1818 CONTINUE
NNE=N
DO 1848 I=1,3
CALL FTAUTO (X,NN,N,0,5,AMEAN,VAR,ACV,AC,PACV,WKAREA)
TEMP=0.0
DO 1828 J=1,M
TEMP=TEMP+AC(J)*AC(J)
1828 CONTINUE
CHI(1)=TEMP
NNE=N-I
DO 1838 K=1,NN
X(K)=X(K+1)-X(K)
1838 CONTINUE
1848 CONTINUE
CALCULATE RATIOS R1 AND R2
R1=CHI(2)/CHI(1)
R2=CHI(3)/CHI(2)
C SELECT NDD
IF (R1.GT.EPSON) GO TO 1858
NDD=1
RETURN
1858 CONTINUE
IF (R2-EPSON) 1868,1868,1878
1868 CONTINUE
NDD=2
RETURN

```

```

2494 WRITE (3,1468) (UK(K),KT,1,LSP)
1368 CONTINUE
1378 READ 1478, (YLAB(J),J=1,8)
IF (NPLT.EQ.0) GO TO 1498
IF (NPLT.GT.2) GO TO 1388
CALL PLOT (U,LSP,9,0,IYR8,YL8)
IF (NPLT.EQ.1) GO TO 1488
1388 CONTINUE
DO 1398 J=1,LSP
U(JT,MP2)=U(JT,MP3)
U(JT,MP3)=U(JT,MP4)
1398 CONTINUE
U(JT,MP4)=U(JT,MP3)
1410 CONTINUE
1420 CALL PLOT (U(1,2)-LSP,7,0,IYR8,YL8)
1430 CONTINUE
1440 FORMAT (16I5)
1450 FORMAT (5F16.6)
1460 FORMAT (10F5.2)
1478 FORMAT (7F5.2)
1488 FORMAT (8A18)
1498 FORMAT (8F18.2)
END

```

```

1878 SUBROUTINE CHECK (N,X,IER)
C-----
C FUNCTION -CHECK VALUES OF PARAMETERS (EITHER AUTOREGRESSIVE OR
C MOVING AVERAGE TYPE) TO SEE IF THEY ARE IN THE
C ADMISSIBLE REGION
C N -NUMBER OF PARAMETERS (EITHER P OR Q OF THE ARIMA
C (P,D,Q) MODEL)
C X -INPUT VECTOR OF LENGTH N
C IER -9, PARAMETERS ARE IN THE ADMISSIBLE REGION
C 139, IF NOT
C-----
DIMENSION X(3)
IF (N.EQ.0) GO TO 1838
ONE=1
IER=0
TP=X(N)
IF (ABS(TP).GT.1.) GO TO 1828
TMP=0.
TP=0.

```



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SUBROUTINE IDENTN (Z,IND,RLIMIT,DISCON,ARF,IPARM,LOGK,NPRINT)  IDE 10
C-----  IDE 20
C FUNCTION  -TO IDENTIFY IP, ID AND IQ OF AN ARMA(LIP, ID, IQ) MODEL  IDE 30
C Z  -INPUT TIME SERIES OF LENGTH N  IDE 40
C IND(I)  I=1, LENGTH OF SERIES  IDE 50
C 2, LENGTH OF PROJECTION (MAXIMUM 600)  IDE 60
C 3, NUMBER OF SETS OF PROJECTION DESIRED (MAXIMUM 10)  IDE 70
C RLIMIT  -MAXIMUM ALLOWABLE RANGE OF PROJECTION IT IS A FUNCTION IDE 80
C OF THE RANGE OF THE CALIBRATION SERIES  IDE 90
C DISCON  -DISCOUNT FACTOR (0.10 TO 0.50) IN COMPUTING DISCOUNTED IDE 100
C SIGNIFICANT LEVEL (AFAL)  IDE 110
C ARF  -OUTPUT VECTOR GIVE FIRST FIVE HIGHEST DISCOUNTED IDE 120
C SIGNIFICANCE LEVELS  IDE 130
C IPARM  -OUTPUT MATRIX GIVES THE P, Q, AND Q FOR THE TOP FIVE IDE 140
C ARMA(P, D, Q) MODELS RANKED DECREASINGLY BY ARF VALUES IDE 150
C LOGK  K=1, LOG10 TRANSFORMATION OF Z IS DESIRED IDE 160
C NPRINT  N=0, NO PRINT OUT IS DESIRED IDE 170
C N=1, PRINT OUT IS DESIRED IDE 180
C-----  IDE 190
C DIMENSION Z(1, J), ALPHA(2), DARPS(5), PHAS(5), FCST(3, 750),  IDE 210
C 1 SIM(750, 5, J), MK(1500), IPDQ(7)  IDE 220
C DIMENSION ZLAST(1, 14), NAMES(14), NP(14), NQ(14), MD(14)  IDE 230
C DIMENSION ALFA(14), ARF(5), RANGEC(14), RANGEP(14),  IDE 240
C IIND(3), IPARM(5, 3)  IDE 250
C DOUBLE PRECISION SEED  IDE 260
C DATA NP/3, 3, 1, 2, 3, 0, 1, 2, 3, 0, 1, 2, 3, 0, 1, 2, 3, 0, 1,  IDE 270
C DATA NQ/3, 2, 3, 1, 2, 3, 0, 2, 1, 0, 2, 1, 0, 2, 1, 0,  IDE 280
C FOR PRINCIPLE COMPONENT ANALYSIS, NO ALSO WILL BE USED  IDE 290
C LOGK=0  IDE 310
C NIND(I)  IDE 320
C LPROJ=IND(2)  IDE 330
C NSET=IND(3)  IDE 340
C CALL FINDO (Z, N, 24, NDD, 1, 0, 40)  IDE 350
C SEARCH FOR OPTIMAL ID AS NDD  IDE 360
C INITIALIZE PARAMETERS TO BE USED IN GTSIMP  IDE 370
C I=750  IDE 380
C IPDQ(1)=1  IDE 390
C IPDQ(2)=N  IDE 400
C IPDQ(3)=LPROJ  IDE 410
C DO 1000 I=1, 14  IDE 420
C SEED=0.15678943256708  IDE 430
C ALPHA(1)=0.0  IDE 440
C ALPHA(2)=0.0  IDE 450
C IPDQ(3)=NP(I)  IDE 460
C IPDQ(4)=NQ(I)  IDE 470
C IPDQ(5)=NDD  IDE 480
C DO 1010 I=1, 5  IDE 490
C DARPS(IN)=0.0  IDE 500
C PHAS(IN)=0.0  IDE 510
C 1010 CONTINUE  IDE 520
C PHAC=0.0  IDE 530
C FCST(2, LPROJ)=0.0  IDE 540
C ZLAST(1, 1)=Z(N)  IDE 550
C DO 1020 I=1, NSET  IDE 560
C ZLAST(I+1, I)=I  IDE 570
C 1020 CONTINUE  IDE 580
C CALL GTSIMP (Z, IPOQ, SEED, ALPHA, DARPS, PHAS, PHAC, MNV, FCST,  IDE 590
C IER)  IDE 610
C-----  IDE 618
SUBROUTINE GTSIMP (Z, IPOQ, SEED, ALPHA, DARPS, PHAS, PHAC, MNV, FCST, SIM, GTS  IH
C 1 IS, MK, IER)  GTS 20
C FUNCTION  -NON-SEASONAL ARIMA STOCHASTIC MODEL ANALYSIS FOR A  GTS 30
C SINGLE TIME SERIES  GTS 40
C -INPUT VECTOR OF LENGTH IPOQ(2) CONTAINING THE  GTS 50
C TIME SERIES  GTS 60
C -INPUT/OUTPUT VECTOR OF LENGTH 7. IPOB(I) CONTAINS,  GTS 70
C GTS 80
C-----  GTS 88
ALFA(I)=ALPHA(I)  IDE 620
IF (IER, GT, 0) ALFA(I)=LE, 0, 0) GO TO 1000  IDE 630
COMPUTE THE RANGE OF PROJECT AS RANGEP  IDE 640
XMIN=SIM(1, 1)  IDE 650
XMAX=SIM(1, 1)  IDE 660
RANGEP(I)=0.0  IDE 670
DO 1040 J=1, NSET  IDE 680
DO 1030 I=1, LPROJ  IDE 690
IF (SIM(I, J), GT, XMAX) XMAX=SIM(I, J)  IDE 700
IF (SIM(I, J), LT, XMIN) XMIN=SIM(I, J)  IDE 710
RPM=MAX(XMIN  IDE 720
RPM=MAX(XMIN  IDE 730
IF (KP, GT, RANGEP(I)) RANGEP(I)=RPM  IDE 740
IF (CHANGE(I)) LE, RLIMIT) GO TO 1040  IDE 750
ALFA(I)=0.0  IDE 760
GO TO 1000  IDE 770
1040 CONTINUE  IDE 780
IF (LOGK, EQ, 0) GO TO 1060  IDE 790
FCST(2, LPROJ)=EXP(FCST(2, LPROJ))  IDE 800
SIM(LPROJ, J)=EXP(SIM(LPROJ, J))  IDE 810
DO 1050 J=1, NSET  IDE 820
SIM(LPROJ, J)=EXP(SIM(LPROJ, J))  IDE 830
DO 1070 I=1, NSET  IDE 840
ZLAST(I+1, I)=SIM(LPROJ, I)  IDE 850
CONTINUE  IDE 860
CONTINUE  IDE 870
J=NSET+1  IDE 880
SEARCH HIGHEST ALFA VALUE AND STORE ITS LOCATION  IDE 890
DISCOUNTED SIGNIFICANCE LEVEL  IDE 900
DO 1090 I=1, 14  IDE 910
ALFA(I)=ALFA(I)*(1.0-DISCON)**(NP(I)*NQ(I)-1)  IDE 920
CONTINUE  IDE 930
RANK ALFA(I) FROM MAX TO MIN AND STORE IN AFA  IDE 940
ASALF(A)  IDE 970
DO 1110 I=1, 5  IDE 980
DO 1100 J=1, 14  IDE 990
IF (ALFA(J), LT, A) GO TO 1100  IDE 1000
A=ALFA(J)  IDE 1010
CONTINUE  IDE 1020
ARF(I)=A  IDE 1030
ALFA(I)=A  IDE 1040
IPARM(1, 1)=NP(I)  IDE 1050
IPARM(1, 2)=NDD  IDE 1060
IPARM(1, 3)=NQ(I)  IDE 1070
CONTINUE  IDE 1080
RETURN  IDE 1090
END  IDE 1100
1110 CONTINUE  IDE 1110
END  IDE 1120
SUBROUTINE GTSIMP (Z, IPOQ, SEED, ALPHA, DARPS, PHAS, PHAC, MNV, FCST, SIM, GTS  IH
C 1 IS, MK, IER)  GTS 20
C FUNCTION  -NON-SEASONAL ARIMA STOCHASTIC MODEL ANALYSIS FOR A  GTS 30
C SINGLE TIME SERIES  GTS 40
C -INPUT VECTOR OF LENGTH IPOQ(2) CONTAINING THE  GTS 50
C TIME SERIES  GTS 60
C -INPUT/OUTPUT VECTOR OF LENGTH 7. IPOB(I) CONTAINS,  GTS 70
C GTS 80

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C 1. I=1, IPDQ(I)=1 IS REQUIRED
C 2. INPUT LENGTH OF TIME SERIES Z.
C 3. NUMBER OF AUTOREGRESSIVE PARAMETERS IN THE
C 4. MODEL
C 5. NUMBER OF MOVING AVERAGE PARAMETERS IN THE MODEL
C 6. DEGREE OF DIFFERENCING OPERATION REQUIRED TO
C 7. OBTAIN A STATIONARY TIME SERIES
C 8. FORECASTING CONTROL PARAMETER.
C 9. FORECASTS UP TO IPDQ(6) STEPS IN ADVANCE ARE
C 10. CALCULATED. IPDQ(6) MUST BE POSITIVE
C 11. I=7, SIMULATION OPTION.
C 12. IPDQ(7) LESS THAN OR EQUAL TO ZERO IMPLIES
C 13. SIMULATION NOT DESIRED
C 14. IPDQ(7) POSITIVE IMPLIES IPDQ(7) SIMULATIONS OF
C 15. THE TIME SERIES UP TO IPDQ(6) STEPS IN ADVANCE
C 16. ARE DESIRED
C 17. INPUT DOUBLE PRECISION NUMBER. SEED IS USED TO
C 18. GENERATE THE SIMULATED TIME SERIES
C 19. INPUT/OUTPUT VECTOR OF LENGTH 2. ALPHA CONTAINS, WHEN
C 20. I=1, INPUT MINIMUM SIGNIFICANCE LEVEL OF THE MODEL
C 21. IN THE EXCLUSIVE RANGE (0,1), ON OUTPUT, THE
C 22. ESTIMATED SIGNIFICANCE LEVEL IS RETURNED.
C 23. I=2, INPUT VALUE IN THE EXCLUSIVE RANGE (0,1)
C 24. USED FOR COMPUTING  $1/(1-\alpha)$  PERCENT
C 25. PROBABILITY LIMITS FOR THE FORECASTS.
C 26. OUTPUT VECTOR OF LENGTH IPDQ(3)*IPDQ(5) CONTAINING
C 27. ESTIMATES OF THE AUTOREGRESSIVE PARAMETERS.
C 28. OUTPUT VECTOR OF LENGTH IPDQ(4) CONTAINING ESTIMATES
C 29. OF THE MOVING AVERAGE PARAMETERS
C 30. OUTPUT ESTIMATE OF THE TREND FACTOR
C 31. OUTPUT ESTIMATE OF THE WHITE NOISE VARIANCE
C 32. OUTPUT MATRIX OF DIMENSION 3 BY IPDQ(6)
C 33. FCST(I,J) FOR LEAD TIMES J=1,2,*,*,*,IPDQ(6),
C 34. CONTAINS, WHEN
C 35. I=1, THE WEIGHTS FOR THE WEIGHTED SUM OF SHOCKS
C 36. THAT GIVES THE FORECAST ERROR
C 37. I=2, THE FORECASTS
C 38. I=3, THE CORRESPONDING DEVIATION FROM EACH
C 39. FORECAST FOR THE  $1/(1-\alpha)$  PERCENT
C 40. PROBABILITY LIMITS
C 41. FORECAST FOR THE  $1/(1-\alpha)$  PERCENT
C 42. OUTPUT MATRIX OF DIMENSION IPDQ(6) BY IPDQ(7)
C 43. DEFINED ONLY WHEN IPDQ(7) IS POSITIVE. SIM(J,I)
C 44. FOR LEAD TIMES J=1,2, IPDQ(6) CONTAINS THE
C 45. RESULTS OF THE IPDQ(7) SIMULATIONS
C 46. I=8, FIRST DIMENSION OF SIM AS DIMENSIONED IN THE
C 47. CALLING PROGRAM.
C 48. CALLABLE LENGTH DEFINED IN THE PROGRAMMING NOTES
C 49. ERROR PARAMETER.
C 50. TERMINAL ERROR. I=20*
C 51. OR FTARPS INDICATES A TERMINAL ERROR OCCURRED IN FTARPS
C 52. WAS OUTPUT RANGE
C 53. INDICATES ALPHA(I), I=1,2 WAS NOT IN THE
C 54. EXCLUSIVE RANGE (0,1).
C 55. WARNING SIGNAL B*EMPHASIS.
C 56. INDICATES THE MODEL DID NOT PASS THE
C 57. ALPHA(I) SIGNIFICANCE TEST. THE NEED FOR MORE
C 58. PARAMETERS IS INDICATED.
C 59. I=1, THIS ROUTINE IS OBTAINED BY MODIFYING THE INBL
C 60. ROUTINE FTIMP.
C 61. I=2, WK IS A WORK VECTOR OF LENGTH AT LEAST THE MAXIMUM OF
C 62.

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C 4. MODEL
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C 18. GENERATE THE SIMULATED TIME SERIES
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C 21. IN THE EXCLUSIVE RANGE (0,1), ON OUTPUT, THE
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C 41. FORECAST FOR THE  $1/(1-\alpha)$  PERCENT
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C 20. I=1, INPUT MINIMUM SIGNIFICANCE LEVEL OF THE MODEL
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C 23. I=2, INPUT VALUE IN THE EXCLUSIVE RANGE (0,1)
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C 36. THAT GIVES THE FORECAST ERROR
C 37. I=2, THE FORECASTS
C 38. I=3, THE CORRESPONDING DEVIATION FROM EACH
C 39. FORECAST FOR THE  $1/(1-\alpha)$  PERCENT
C 40. PROBABILITY LIMITS
C 41. FORECAST FOR THE  $1/(1-\alpha)$  PERCENT
C 42. OUTPUT MATRIX OF DIMENSION IPDQ(6) BY IPDQ(7)
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C 60. ROUTINE FTIMP.
C 61. I=2, WK IS A WORK VECTOR OF LENGTH AT LEAST THE MAXIMUM OF
C 62.

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1200 CONTINUE
1210 CONTINUE
1220 CALL FCST (Z,MK,MK(N+1),PHAC,ALPHA(2),IPDQ(2),DARPS,FCST,MNV,
1 IERR)
C SIMULATIONS
IF (IPDQ(7).NE.0) GO TO 1220
MIPDQ(7)
J=TP+10
KMAX(J)=1
DO 1220 I=1,M
PHACTEP=MEAN(I)
IF (IPDQ(8).NE.0) PHAC=PHAC*(1.0-0.2*IPDQ(8)/IPDQ(2)*0.1)
CALL FCST (DARPS,PHAC,Z(N+K+1),MNV,SEED,7,IPDQ(8),SIM(1),
1 IERR)
1220 CONTINUE
1230 RETURN
1230 CONTINUE
ALPHA(1)=1.0
RETURN
1240 CONTINUE
ALPHA(1)=2.0
RETURN
1250 CONTINUE
ALPHA(1)=3.0
RETURN
1260 CONTINUE
ALPHA(1)=4.0
RETURN
1270 CONTINUE
ALPHA(1)=5.0
RETURN
END
SUBROUTINE PLOT (G,N,MV,IFCTR,IBEGIN,YLABEL)
C FUNCTION - TO PLOT UP TO 9 CURVES IN ONE CHART
C DATA LTR(1)H1,1H2,1H3,1H4,1H5,1H6,1H7,1H8,1H9/
C N - LENGTH OF EACH VARIABLE IN G(I,J)
C NV - NUMBER OF VARIABLES IN G(I,J) TO BE PLOTTED
C IFCTR - SCALING FACTOR
C IFRGIN - BEGINNING DATE OF THE RECORD (USED IN X-AXIS)
C YLABEL - LABEL TO BE USED IN THE Y-AXIS OF THE PLOT
C NOTE 1. DIMENSIONS OF THE PARAMETERS IPLOT AND ISVAE (CONTROL
2. THE PRINTING SPACE) ARE NOT CHANGEABLE
3. DIMENSION OF G AND THE FIRST IF STATEMENT SHOULD BE
CHANGED ACCORDING TO THE SIZE OF THE MATRIX PLOTTED
DIMENSION IPLOT(121),LTR(10),ISVAE(121),SCALE(13),G(259,9)
DIMENSION YLABEL(8)
DATA LTR/1H1,1H2,1H3,1H4,1H5,1H6,1H7,1H8,1H9/
DATA ISVAE/1H1,1H2,1H3,1H4,1H5,1H6,1H7,1H8,1H9/
WRITE (6,1200) (YLABEL(J),J=1,8)
IF (N.GT.259.OR.NV.GT.9) GO TO 1198
TMP=10.**IFCTR
DO 1910 I=1,121
ISVAE(I)=IBLANK
IPLOT(I)=IBLANK
1910 CONTINUE
WRITE (6,1220) TMP
END

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GTS133X
GTS1340
GTS1350
GTS1360
GTS1370
GTS1380
GTS1390
GTS1400
GTS1410
GTS1420
GTS1430
GTS1440
GTS1450
GTS1460
GTS1470
GTS1480
GTS1490
GTS1500
GTS1510
GTS1520
GTS1530
GTS1540
GTS1550
GTS1560
GTS1570
GTS1580
GTS1590
GTS1600
GTS1610
GTS1620
GTS1630
GTS1640
GTS1650
GTS1660
GTS1670
GTS1680
GTS1690
GTS1700
GTS1710
GTS1720
GTS1730
GTS1740
GTS1750
GTS1760
GTS1770
GTS1780
GTS1790
GTS1800
GTS1810
GTS1820
GTS1830
GTS1840
GTS1850
GTS1860
GTS1870
GTS1880
GTS1890
GTS1900
GTS1910
GTS1920
GTS1930
GTS1940
TEMPMAC/MBAR
IF (IER.GT.8) GO TO 1230
CALL CHECK (IPDQ(3),DARPS(1),IER)
IF (IER.GT.8) GO TO 1240
IF (IPDQ(4).EQ.0) GO TO 1800
1870 CONTINUE
CALL FMAPS (MK(N+1),DARPS,IPDQ(3),IPDQ(4),PHAS,MNV,MK(J),IER)
IF (IER.GT.8) GO TO 1250
IF (IPDQ(3).EQ.0) PHAC=MBAR
1880 CONTINUE
CALL CHECK (IPDQ(4),PHAS(1),IER)
IF (IER.GT.8) GO TO 1260
FIND ONE-STEP FORECAST ERROR
DO 1890 I=1,IR
MK(N+1)=0.0
1890 CONTINUE
DO 1900 I=1,M
MK(I)=MK(I)+MBAR
1900 CONTINUE
DO 1910 I=1,M
MK(I)=MK(I)+MBAR
1910 CONTINUE
DO 1920 I=1,M
APL=APL+I*IR,N
MK(NPT)=MK(I)
DO 1920 I=1,M
MK(NPT)=MK(NPT)+DARPS(J)*MK(I-J)
1920 CONTINUE
IF (I.OE.0) GO TO 1950
1930 CONTINUE
DO 1940 I=1,10
MK(NPT)=MK(NPT)+PHAS(J)*MK(NPT-I)
1940 CONTINUE
1950 CONTINUE
C FIND COVARIANCE OF ERRORS
CALL FTAUTO (MK(N+1),N,120,10,15,MK(N2421),MK(N2522),MK(N2623),
1 MK(N2724),MK,N)
C FIND ERROR SIGNIFICANCE STATISTIC
TEMP=0
DO 1960 I=1,120
TEMP=TEMP+I**2
1960 CONTINUE
TEMP=TEMP/120
MNV=IPDQ(5)
MNT=IPDQ(6)
MNT2=IP-10
CALL MCHN (I,MV,CHI,IER)
CALL MCHN (I,MV,CHI,IER)
IF ((1.0-CHI).GE.ALPHA(1)) GO TO 1170
GO TO 1270
FORECASTS
1170 CONTINUE
ALPHA(1)=1.0-CHI
C LINEAR DECAY OF THE PHAC WHEN ID GT 8
IF (ID.GT.8) PHAC=PHAC*(1.0-0.2*IPDQ(4)/IPDQ(2)*0.1)
DO 1180 I=1,121
PHAC(I)=PHAS(I)
1180 CONTINUE
1190 CONTINUE
DO 1200 I=1,10
IPDQ(I)=IBLANK
1200 CONTINUE
DO 1210 I=1,10
IPDQ(I)=IBLANK
1210 CONTINUE
DO 1220 I=1,10
IPDQ(I)=IBLANK
1220 CONTINUE
DO 1230 I=1,10
IPDQ(I)=IBLANK
1230 CONTINUE
DO 1240 I=1,10
IPDQ(I)=IBLANK
1240 CONTINUE
DO 1250 I=1,10
IPDQ(I)=IBLANK
1250 CONTINUE
DO 1260 I=1,10
IPDQ(I)=IBLANK
1260 CONTINUE
DO 1270 I=1,10
IPDQ(I)=IBLANK
1270 CONTINUE
DO 1280 I=1,10
IPDQ(I)=IBLANK
1280 CONTINUE
DO 1290 I=1,10
IPDQ(I)=IBLANK
1290 CONTINUE
DO 1300 I=1,10
IPDQ(I)=IBLANK
1300 CONTINUE
DO 1310 I=1,10
IPDQ(I)=IBLANK
1310 CONTINUE
DO 1320 I=1,10
IPDQ(I)=IBLANK
1320 CONTINUE
DO 1330 I=1,10
IPDQ(I)=IBLANK
1330 CONTINUE
DO 1340 I=1,10
IPDQ(I)=IBLANK
1340 CONTINUE
DO 1350 I=1,10
IPDQ(I)=IBLANK
1350 CONTINUE
DO 1360 I=1,10
IPDQ(I)=IBLANK
1360 CONTINUE
DO 1370 I=1,10
IPDQ(I)=IBLANK
1370 CONTINUE
DO 1380 I=1,10
IPDQ(I)=IBLANK
1380 CONTINUE
DO 1390 I=1,10
IPDQ(I)=IBLANK
1390 CONTINUE
DO 1400 I=1,10
IPDQ(I)=IBLANK
1400 CONTINUE
DO 1410 I=1,10
IPDQ(I)=IBLANK
1410 CONTINUE
DO 1420 I=1,10
IPDQ(I)=IBLANK
1420 CONTINUE
DO 1430 I=1,10
IPDQ(I)=IBLANK
1430 CONTINUE
DO 1440 I=1,10
IPDQ(I)=IBLANK
1440 CONTINUE
DO 1450 I=1,10
IPDQ(I)=IBLANK
1450 CONTINUE
DO 1460 I=1,10
IPDQ(I)=IBLANK
1460 CONTINUE
DO 1470 I=1,10
IPDQ(I)=IBLANK
1470 CONTINUE
DO 1480 I=1,10
IPDQ(I)=IBLANK
1480 CONTINUE
DO 1490 I=1,10
IPDQ(I)=IBLANK
1490 CONTINUE
DO 1500 I=1,10
IPDQ(I)=IBLANK
1500 CONTINUE
DO 1510 I=1,10
IPDQ(I)=IBLANK
1510 CONTINUE
DO 1520 I=1,10
IPDQ(I)=IBLANK
1520 CONTINUE
DO 1530 I=1,10
IPDQ(I)=IBLANK
1530 CONTINUE
DO 1540 I=1,10
IPDQ(I)=IBLANK
1540 CONTINUE
DO 1550 I=1,10
IPDQ(I)=IBLANK
1550 CONTINUE
DO 1560 I=1,10
IPDQ(I)=IBLANK
1560 CONTINUE
DO 1570 I=1,10
IPDQ(I)=IBLANK
1570 CONTINUE
DO 1580 I=1,10
IPDQ(I)=IBLANK
1580 CONTINUE
DO 1590 I=1,10
IPDQ(I)=IBLANK
1590 CONTINUE
DO 1600 I=1,10
IPDQ(I)=IBLANK
1600 CONTINUE
DO 1610 I=1,10
IPDQ(I)=IBLANK
1610 CONTINUE
DO 1620 I=1,10
IPDQ(I)=IBLANK
1620 CONTINUE
DO 1630 I=1,10
IPDQ(I)=IBLANK
1630 CONTINUE
DO 1640 I=1,10
IPDQ(I)=IBLANK
1640 CONTINUE
DO 1650 I=1,10
IPDQ(I)=IBLANK
1650 CONTINUE
DO 1660 I=1,10
IPDQ(I)=IBLANK
1660 CONTINUE
DO 1670 I=1,10
IPDQ(I)=IBLANK
1670 CONTINUE
DO 1680 I=1,10
IPDQ(I)=IBLANK
1680 CONTINUE
DO 1690 I=1,10
IPDQ(I)=IBLANK
1690 CONTINUE
DO 1700 I=1,10
IPDQ(I)=IBLANK
1700 CONTINUE
DO 1710 I=1,10
IPDQ(I)=IBLANK
1710 CONTINUE
DO 1720 I=1,10
IPDQ(I)=IBLANK
1720 CONTINUE
DO 1730 I=1,10
IPDQ(I)=IBLANK
1730 CONTINUE
DO 1740 I=1,10
IPDQ(I)=IBLANK
1740 CONTINUE
DO 1750 I=1,10
IPDQ(I)=IBLANK
1750 CONTINUE
DO 1760 I=1,10
IPDQ(I)=IBLANK
1760 CONTINUE
DO 1770 I=1,10
IPDQ(I)=IBLANK
1770 CONTINUE
DO 1780 I=1,10
IPDQ(I)=IBLANK
1780 CONTINUE
DO 1790 I=1,10
IPDQ(I)=IBLANK
1790 CONTINUE
DO 1800 I=1,10
IPDQ(I)=IBLANK
1800 CONTINUE
DO 1810 I=1,10
IPDQ(I)=IBLANK
1810 CONTINUE
DO 1820 I=1,10
IPDQ(I)=IBLANK
1820 CONTINUE
DO 1830 I=1,10
IPDQ(I)=IBLANK
1830 CONTINUE
DO 1840 I=1,10
IPDQ(I)=IBLANK
1840 CONTINUE
DO 1850 I=1,10
IPDQ(I)=IBLANK
1850 CONTINUE
DO 1860 I=1,10
IPDQ(I)=IBLANK
1860 CONTINUE
DO 1870 I=1,10
IPDQ(I)=IBLANK
1870 CONTINUE
DO 1880 I=1,10
IPDQ(I)=IBLANK
1880 CONTINUE
DO 1890 I=1,10
IPDQ(I)=IBLANK
1890 CONTINUE
DO 1900 I=1,10
IPDQ(I)=IBLANK
1900 CONTINUE
DO 1910 I=1,10
IPDQ(I)=IBLANK
1910 CONTINUE
DO 1920 I=1,10
IPDQ(I)=IBLANK
1920 CONTINUE
DO 1930 I=1,10
IPDQ(I)=IBLANK
1930 CONTINUE
DO 1940 I=1,10
IPDQ(I)=IBLANK
1940 CONTINUE
DO 1950 I=1,10
IPDQ(I)=IBLANK
1950 CONTINUE
DO 1960 I=1,10
IPDQ(I)=IBLANK
1960 CONTINUE
DO 1970 I=1,10
IPDQ(I)=IBLANK
1970 CONTINUE
DO 1980 I=1,10
IPDQ(I)=IBLANK
1980 CONTINUE
DO 1990 I=1,10
IPDQ(I)=IBLANK
1990 CONTINUE
DO 2000 I=1,10
IPDQ(I)=IBLANK
2000 CONTINUE
DO 2010 I=1,10
IPDQ(I)=IBLANK
2010 CONTINUE
DO 2020 I=1,10
IPDQ(I)=IBLANK
2020 CONTINUE
DO 2030 I=1,10
IPDQ(I)=IBLANK
2030 CONTINUE
DO 2040 I=1,10
IPDQ(I)=IBLANK
2040 CONTINUE
DO 2050 I=1,10
IPDQ(I)=IBLANK
2050 CONTINUE
DO 2060 I=1,10
IPDQ(I)=IBLANK
2060 CONTINUE
DO 2070 I=1,10
IPDQ(I)=IBLANK
2070 CONTINUE
DO 2080 I=1,10
IPDQ(I)=IBLANK
2080 CONTINUE
DO 2090 I=1,10
IPDQ(I)=IBLANK
2090 CONTINUE
DO 2100 I=1,10
IPDQ(I)=IBLANK
2100 CONTINUE
DO 2110 I=1,10
IPDQ(I)=IBLANK
2110 CONTINUE
DO 2120 I=1,10
IPDQ(I)=IBLANK
2120 CONTINUE
DO 2130 I=1,10
IPDQ(I)=IBLANK
2130 CONTINUE
DO 2140 I=1,10
IPDQ(I)=IBLANK
2140 CONTINUE
DO 2150 I=1,10
IPDQ(I)=IBLANK
2150 CONTINUE
DO 2160 I=1,10
IPDQ(I)=IBLANK
2160 CONTINUE
DO 2170 I=1,10
IPDQ(I)=IBLANK
2170 CONTINUE
DO 2180 I=1,10
IPDQ(I)=IBLANK
2180 CONTINUE
DO 2190 I=1,10
IPDQ(I)=IBLANK
2190 CONTINUE
DO 2200 I=1,10
IPDQ(I)=IBLANK
2200 CONTINUE
DO 2210 I=1,10
IPDQ(I)=IBLANK
2210 CONTINUE
DO 2220 I=1,10
IPDQ(I)=IBLANK
2220 CONTINUE
DO 2230 I=1,10
IPDQ(I)=IBLANK
2230 CONTINUE
DO 2240 I=1,10
IPDQ(I)=IBLANK
2240 CONTINUE
DO 2250 I=1,10
IPDQ(I)=IBLANK
2250 CONTINUE
DO 2260 I=1,10
IPDQ(I)=IBLANK
2260 CONTINUE
DO 2270 I=1,10
IPDQ(I)=IBLANK
2270 CONTINUE
DO 2280 I=1,10
IPDQ(I)=IBLANK
2280 CONTINUE
DO 2290 I=1,10
IPDQ(I)=IBLANK
2290 CONTINUE
DO 2300 I=1,10
IPDQ(I)=IBLANK
2300 CONTINUE
DO 2310 I=1,10
IPDQ(I)=IBLANK
2310 CONTINUE
DO 2320 I=1,10
IPDQ(I)=IBLANK
2320 CONTINUE
DO 2330 I=1,10
IPDQ(I)=IBLANK
2330 CONTINUE
DO 2340 I=1,10
IPDQ(I)=IBLANK
2340 CONTINUE
DO 2350 I=1,10
IPDQ(I)=IBLANK
2350 CONTINUE
DO 2360 I=1,10
IPDQ(I)=IBLANK
2360 CONTINUE
DO 2370 I=1,10
IPDQ(I)=IBLANK
2370 CONTINUE
DO 2380 I=1,10
IPDQ(I)=IBLANK
2380 CONTINUE
DO 2390 I=1,10
IPDQ(I)=IBLANK
2390 CONTINUE
DO 2400 I=1,10
IPDQ(I)=IBLANK
2400 CONTINUE
DO 2410 I=1,10
IPDQ(I)=IBLANK
2410 CONTINUE
DO 2420 I=1,10
IPDQ(I)=IBLANK
2420 CONTINUE
DO 2430 I=1,10
IPDQ(I)=IBLANK
2430 CONTINUE
DO 2440 I=1,10
IPDQ(I)=IBLANK
2440 CONTINUE
DO 2450 I=1,10
IPDQ(I)=IBLANK
2450 CONTINUE
DO 2460 I=1,10
IPDQ(I)=IBLANK
2460 CONTINUE
DO 2470 I=1,10
IPDQ(I)=IBLANK
2470 CONTINUE
DO 2480 I=1,10
IPDQ(I)=IBLANK
2480 CONTINUE
DO 2490 I=1,10
IPDQ(I)=IBLANK
2490 CONTINUE
DO 2500 I=1,10
IPDQ(I)=IBLANK
2500 CONTINUE
DO 2510 I=1,10
IPDQ(I)=IBLANK
2510 CONTINUE
DO 2520 I=1,10
IPDQ(I)=IBLANK
2520 CONTINUE
DO 2530 I=1,10
IPDQ(I)=IBLANK
2530 CONTINUE
DO 2540 I=1,10
IPDQ(I)=IBLANK
2540 CONTINUE
DO 2550 I=1,10
IPDQ(I)=IBLANK
2550 CONTINUE
DO 2560 I=1,10
IPDQ(I)=IBLANK
2560 CONTINUE
DO 2570 I=1,10
IPDQ(I)=IBLANK
2570 CONTINUE
DO 2580 I=1,10
IPDQ(I)=IBLANK
2580 CONTINUE
DO 2590 I=1,10
IPDQ(I)=IBLANK
2590 CONTINUE
DO 2600 I=1,10
IPDQ(I)=IBLANK
2600 CONTINUE
DO 2610 I=1,10
IPDQ(I)=IBLANK
2610 CONTINUE
DO 2620 I=1,10
IPDQ(I)=IBLANK
2620 CONTINUE
DO 2630 I=1,10
IPDQ(I)=IBLANK
2630 CONTINUE
DO 2640 I=1,10
IPDQ(I)=IBLANK
2640 CONTINUE
DO 2650 I=1,10
IPDQ(I)=IBLANK
2650 CONTINUE
DO 2660 I=1,10
IPDQ(I)=IBLANK
2660 CONTINUE
DO 2670 I=1,10
IPDQ(I)=IBLANK
2670 CONTINUE
DO 2680 I=1,10
IPDQ(I)=IBLANK
2680 CONTINUE
DO 2690 I=1,10
IPDQ(I)=IBLANK
2690 CONTINUE
DO 2700 I=1,10
IPDQ(I)=IBLANK
2700 CONTINUE
DO 2710 I=1,10
IPDQ(I)=IBLANK
2710 CONTINUE
DO 2720 I=1,10
IPDQ(I)=IBLANK
2720 CONTINUE
DO 2730 I=1,10
IPDQ(I)=IBLANK
2730 CONTINUE
DO 2740 I=1,10
IPDQ(I)=IBLANK
2740 CONTINUE
DO 2750 I=1,10
IPDQ(I)=IBLANK
2750 CONTINUE
DO 2760 I=1,10
IPDQ(I)=IBLANK
2760 CONTINUE
DO 2770 I=1,10
IPDQ(I)=IBLANK
2770 CONTINUE
DO 2780 I=1,10
IPDQ(I)=IBLANK
2780 CONTINUE
DO 2790 I=1,10
IPDQ(I)=IBLANK
2790 CONTINUE
DO 2800 I=1,10
IPDQ(I)=IBLANK
2800 CONTINUE
DO 2810 I=1,10
IPDQ(I)=IBLANK
2810 CONTINUE
DO 2820 I=1,10
IPDQ(I)=IBLANK
2820 CONTINUE
DO 2830 I=1,10
IPDQ(I)=IBLANK
2830 CONTINUE
DO 2840 I=1,10
IPDQ(I)=IBLANK
2840 CONTINUE
DO 2850 I=1,10
IPDQ(I)=IBLANK
2850 CONTINUE
DO 2860 I=1,10
IPDQ(I)=IBLANK
2860 CONTINUE
DO 2870 I=1,10
IPDQ(I)=IBLANK
2870 CONTINUE
DO 2880 I=1,10
IPDQ(I)=IBLANK
2880 CONTINUE
DO 2890 I=1,10
IPDQ(I)=IBLANK
2890 CONTINUE
DO 2900 I=1,10
IPDQ(I)=IBLANK
2900 CONTINUE
DO 2910 I=1,10
IPDQ(I)=IBLANK
2910 CONTINUE
DO 2920 I=1,10
IPDQ(I)=IBLANK
2920 CONTINUE
DO 2930 I=1,10
IPDQ(I)=IBLANK
2930 CONTINUE
DO 2940 I=1,10
IPDQ(I)=IBLANK
2940 CONTINUE
DO 2950 I=1,10
IPDQ(I)=IBLANK
2950 CONTINUE
DO 2960 I=1,10
IPDQ(I)=IBLANK
2960 CONTINUE
DO 2970 I=1,10
IPDQ(I)=IBLANK
2970 CONTINUE
DO 2980 I=1,10
IPDQ(I)=IBLANK
2980 CONTINUE
DO 2990 I=1,10
IPDQ(I)=IBLANK
2990 CONTINUE
DO 3000 I=1,10
IPDQ(I)=IBLANK
3000 CONTINUE

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1120 CONTINUE
STEP=STEP+1
TEMP=ITP
START=TEMP*STEP
LX=1
DO 1130 L=1,13
ISAVE(LX)=IPER
IPLOT(LX)=IPER
LX=LX+1
1130 CONTINUE
KOEY=IREGIN
DO 1140 K=1,N
IF (K.GT.1.AND.K.LT.N) GO TO 1150
DO 1140 L=1,121
IPLOT(L)=IPER
1140 CONTINUE
1150 DO 1160 I=1,NV
TEMP=K(I)
ITMP=(TEMP-START)/STEP+.5
LX=ITMP+1STEP+1
IF (LX.GT.121) LX=121
IF (LX.LT.1) LX=1
IPLOT(LX)=LTR(I)
1160 CONTINUE
WRITE (6,1260) KOEX,IPLOT
KOEY=1+KOEY
LACR=LAG+1
DO 1170 L=1,121
IPLOT(L)=ISAVE(L)
1170 CONTINUE
1180 CONTINUE
WRITE (6,1270)
1190 RETURN
WRITE (6,1280)
RETURN
1200 FORMAT (B10)
1210 FORMAT (10X,20HQUANTITIES SCALED BY ,F6.0)
1220 FORMAT (10X)
1230 FORMAT (2X,F6.0,12F10.0)
1240 FORMAT (2X,F6.0,12F10.1)
1250 FORMAT (2X,F6.0,12F10.2)
1260 FORMAT (18,2X,121A1)
1270 FORMAT (10X)
1280 FORMAT (1X,18HDIMENSION EXCEEDED)
END

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1020 CONTINUE
AMIN=999999.
AMAX=999999.
DO 1020 I=1,NV
DO 1020 J=1,N
O(I,J)=O(I,J)/TMP
TEMP=O(I,J)
IF (TEMP.GT.AMAX) AMAX=TEMP
IF (TEMP.LT.AMIN) AMIN=TEMP
1020 CONTINUE
IF (AMAX.LE.AMIN) AMAX=1.01*AMIN
RATIO=(AMAX-AMIN)/TMP
STEP=1.
1030 CONTINUE
IF (RATIO.LE.5) GO TO 1040
IF (RATIO.GT.5) GO TO 1050
IF (RATIO.LE.2.AND.RATIO.GT.1) STEP=2.*STEP
IF (RATIO.LE.1.AND.RATIO.GT.1) STEP=4.*STEP
IF (RATIO.GT.4) STEP=5.*STEP
GO TO 1060
1040 CONTINUE
STEP=STEP*.1
RATIO=RATIO*.18
IF (STEP.LT.1)
IF (STEP.GT.1)
GO TO 1030
1050 CONTINUE
STEP=STEP*.14
RATIO=RATIO*.1
GO TO 1030
1060 CONTINUE
ITP=AMIN/STEP
TEMP=ITP
IF (AMIN.LT.TMP) ITP=ITP-.1
ITMP=AMAX/STEP
TEMP=ITP
ITMP=(ITMP+.01)*STEP
IF (AMAX.GT.TMP) ITP=ITP+.1
ITMP=1
DO 1070 L=1,13
SCALE(L)=0.
LX=1
CONTINUE
TEMP=ITP
TEMP=TEMP*STEP
ITMP=ITP+1-ITP
SCALE(LX)=TEMP
LX=LX+1TEMP
TEMP=TEMP*STEP
IF (LX.LE.13) GO TO 1080
1080 CONTINUE
WRITE (6,1230) SCALE
GO TO 1120
1100 CONTINUE
WRITE (6,1240) SCALE
GO TO 1120
1110 CONTINUE
WRITE (6,1250) SCALE.

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