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- no 2 -

ARSTU

User's Manual

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*Deze software is tot stand gekomen mede dank zij de steun van de
Christelijke Centrale van Houtbewerkers en Bouwvakarbeiders.*

*We thank Mr. H. Pauwels. Without his help this program would never have
been achieved.*

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Abstract

ARSTU is a computer program for the estimation of Auto Regressive Models for Stationary Time Series in the Univariate case.

Two modelling procedures are available : a Minimum AIC procedure and a Quasi Bayesian procedure. In order to evaluate the estimated AR-model the computation of the characteristic roots of the polynomial in the lag operator and an estimation of the spectrum based on the estimated AR-coefficients are included as additional features.

The input data file for ARSTU can also be used for the TSERS, PACK (PACK 1, PACK 2 & Pack 3), MULTISTOCH and MULTITRAN programs.

TABLE OF CONTENTS

1. General Description	2
2. Algorithmic and Computational Aspects	4
3. Description of Input	10
4. Description of Output	14
5. Output Example : ARMFIT	16
6. Output Example : ARBAYS	21
7. Output Example : ARMLE	27
Appendix I : Input Datafile	28
Appendix II : User-Level Flow Chart	30
Appendix III : Subroutine Map	31
References	33

1. GENERAL DESCRIPTION

ARSTU is a computer program for the estimation of

Auto
Regressive Models for
Stationary Time Series in the
Univariate case

Two modelling procedures are available:

1. Minimum AIC Procedure

The method of LEAST SQUARES is applied to estimate the parameters of a particular model and an information criterion is computed by

$$AIC = - 2 \ln \left(\frac{\text{innovation}}{\text{variance}} \right) + 2 \left(\frac{\text{number of}}{\text{parameters}} \right)$$

The model which attains the minimum of the AIC's among various models is selected as the final choice.

2. Quasi Bayesian Procedure

The quasi BAYESIAN procedure is obtained by taking the average of the models, with the parameters estimated by the method of LEAST SQUARES, with respect to the "posterior" probability which is by definition proportional to

$$e^{-\frac{1}{2} AIC} \times (\text{prior probability of the model}).$$

It is expected that this Bayesian type model fitting procedure will reduce the risk of adopting too high order models by the simple minimum AIC procedure.

Apart from the LEAST SQUARES estimation method also the MAXIMUM LIKELIHOOD estimation method is incorporated in the ARSTU program. This feature makes it possible to use the autoregressive coefficients, estimated by Least Squares, as (good) starting values for the exact GAUSSIAN LIKELIHOOD computation of the AR model.

In order to evaluate the estimated AR model two additional features are included in the ARSTU program.

1. Characteristic Roots

If requested the stability of the AR operator can be checked by computing the characteristic roots of the polynomial in the lag operator.

2. Spectrum estimation

To facilitate the understanding of the behaviour of the time series considered the estimated AR coefficients can be used to derive the spectrum of that time series.

The ARSTU program is mainly based on three programs from the TIMSAC-78 computer package, implemented by H. AKAIKE from the Institute of Statistical Mathematics (Tokyo). These three programs are

1. UNIMAR (TIMSAC 78.1.1)

UNIvariate Case of
Minimum AIC Method of
AR Model Fitting

2. UNIBAR (TIMSAC 78.1.2)

UNIvariate Case of
Bayesian Method of
AR Model Fitting

3. EXSAR (TIMSAC 78.5.1)

EXact Maximum Likelihood Method of
Scalar
AR Model Fitting

For more details on the TIMSAC-78 computer package, which is a continuation of the former packages TIMSAC and TIMSAC-74, see [7] and also [8], [5] & [6].

2. ALGORITHMIC AND COMPUTATIONAL ASPECTS

In this section the most important algorithmic and computational aspects of ARSTU will be discussed.

2.1. Subroutine REDATA

This subroutine, which is a TIMSAC-78 subroutine, is used for the loading of the original data and deletion of the mean value.

The data are loaded through the device specified by MT, i.e. through a datafile on disk (MT = 1) or the terminal keyboard (MT = 0).

For further details about the input of the time series data see Section 3.2.

2.2. Subroutine REDUCT

This subroutine, which is a TIMSAC-78 subroutine, first sets up a data matrix X by augmenting the successively shifted original data vector Z and then transforms X into triangular form by the HOUSEHOLDER transformation. This transformation is applied directly to the data matrix to avoid the computation of sample covariances.

This procedure allows a very simple revision of models when a new additional set of data is added to the original one.

For further information on the use of the HOUSEHOLDER transformation in this kind of applications see [12].

2.3. Subroutine ARMFIT

This is the basic subroutine for the UNIvariate case of Minimum AIC method of AR model fitting.

This subroutine fits autoregressive models of successively higher orders by the method of LEAST SQUARES and realized through HOUSEHOLDER's transformation [12].

First the data matrix X is generated where X depends on the model to be fitted. For the fitting of a regression model

$$x_n = a_1 y_{n1} + a_2 y_{n2} + \dots + a_M y_{nM} + \epsilon_n$$

where

ϵ_n is a zero mean white noise process
and

$$y_{nj} = x_{n-j}$$

the X matrix is defined by

$$X = \begin{pmatrix} y_{1,1} & \cdots & y_{1,M} & x_1 \\ y_{2,1} & & y_{2,M} & x_2 \\ \vdots & & & \\ y_{N,1} & \cdots & y_{N,M} & x_N \end{pmatrix}$$

where the observations of M regressors constitute the first M columns and those of the regress and constitute the last column.

By applying the HOUSEHOLDER transformation X is transformed into an upper triangular matrix S

$$S = \begin{pmatrix} S_{1,1} & \cdots & S_{1,M} & S_{1,M+1} \\ & \ddots & & \\ & & \ddots & \\ 0 & & & S_{M,M} & S_{M,M+1} \\ & & & & S_{M+1,M+1} \end{pmatrix}$$

The AIC of the model, defined with the first m regressors, is obtained by

$$AIC(m) = N \ln |\hat{\sigma}^2(m)| + 2(m+1)$$

where

$$\hat{\sigma}^2(m) = \sum_{i=m+1}^{M+1} S_{i,M+1}^2$$

The regression coefficients of the model with m regressors are then given by

$$a_m = S_{m,M+1} / S_{m,m}$$

$$a_i = |S_{i,M+1} - \sum_{j=i-1}^m a_j S_{j,M+1}| / S_{i,i} \quad (i = m-1, \dots, 1)$$

For further details about the LEAST SQUARES method and the AIC procedure see [13] and [3].

2.4. Subroutine ARBAYS

This is the basic subroutine for the UNI variate Bayesian method of AR model fitting (UNIBAR).

This subroutine fits an autoregressive model by a BAYESIAN procedure.

The posterior weight of the m^{th} order model is given by

$$w(m) = C \pi(m) c(m)$$

where the prior weight $\pi(m)$ is given by

$$\pi(m) = \frac{1}{m + 1}$$

$c(m)$ is defined as

$$c(m) = \exp\left\{-\frac{1}{2} |AIC(m) - MAIC|\right\}$$

where MAIC is the minimum of AIC(m) and the normalizing constant is given by

$$C^{-1} = \sum_{m=0}^M \pi(m) c(m)$$

The partial autoregression coefficients are then computed as

$$b^*(m) = b(m) d(m)$$

where $b(m)$ is the least squares estimate and the integrated Bayesian weight $d(m)$ is defined as

$$d(m) = \sum_{i=m}^{M+1} c(i)$$

The autoregressive coefficients of the Bayesian model are then derived as a conversion from the set of partial autoregression coefficients ($b^*(m)$) to the autoregressive coefficients a_i ($i = 1, 2, \dots, M$).

For evaluation purposes the innovation variance of the Bayesian model can be computed as

$$\sigma_B^2 = \frac{1}{N} \sum_{n=1}^N \left(x_n - \sum_{i=1}^M a_i y_{ni} \right)^2$$

For the Bayesian model an equivalent AIC is defined as

$$\text{AICB} = N \ln |\sigma_B^2| + 2 \text{EK}$$

where the equivalent number of parameters EK is given by

$$\text{EK} = \sum_{m=1}^M d(m)^2 + 1$$

For further details about this Bayesian procedure see [4] and [1].

2.5. Subroutine ARMLE

This is the basic subroutine for the EXact maximum likelihood method of Scalar AR-model fitting (EXSAR).

This subroutine produces exact MAXIMUM LIKELIHOOD estimates of the parameters of a scalar AR-model.

Given the data (x_1, x_2, \dots, x_N) the likelihood of the autoregressive model

$$x_n = a_1 x_{n-1} + \dots + a_m x_{n-m} + \varepsilon_n$$

where

$$\varepsilon_n \sim N(0, \sigma^2)$$

is obtained by

$$\begin{aligned} & p(x_1, \dots, x_N) \mid a_1, \dots, a_m, \sigma^2 \\ &= p(x_1, \dots, x_m \mid a_1, \dots, a_m, \sigma^2) \\ & \quad \prod_{i=m+1}^N p(x_i \mid x_{i-1}, \dots, x_{i-m}, a_1, \dots, a_m, \sigma^2) \end{aligned}$$

where

$$\begin{aligned} & p(x_i \mid x_{i-1}, \dots, x_{i-m}, a_1, \dots, a_m, \sigma^2) \\ &= \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{1}{2\sigma^2} (x_n - a_1 x_{n-1} - \dots - a_m x_{n-m})^2 \right\} \end{aligned}$$

and

$$p(x_1, \dots, x_m | a_1, \dots, a_m, \sigma^2)$$

is evaluated by using the theoretical autocovariance function $\rho_0, \rho_1, \dots, \rho_m$ determined by a_1, \dots, a_m and σ^2 .

For the minimization the DAVIDON-FLETCHER-POWELL procedure is used. (See [9] and [11]).

2.6. Subroutine ARMACR

This subroutine is mainly based on the TIMSAC-78 subroutine CHROOT.

The subroutine ARMACR finds the characteristic roots of

- autoregressive operator - all coefficients
- autoregressive operator - subset of coefficients
- AR.MA operators

The output consists of

- real part of the root (R)
- imaginary part of the root (I)
- modulus (SQRT (R ** 2 + I ** 2))
- phase in radii (ARCTAN (I/R))
- phase in degrees

Most of the calculations are performed by subroutine POLYRT. This subroutine computes the real and complex roots of a real polynomial by the NEWTON-RAPHSON iterative technique. The final iterations on each root are performed using the original polynomial rather than the reduced polynomial to avoid accumulated errors in the reduced polynomial.

Floating point overflow may occur for high order polynomials but will not affect the accuracy of the results. If the non-convergence condition occurs the error message

**** NON CONVERGENCE AT POLYRT ****

is given.

2.7. Subroutine ARMASE

This subroutine is mainly based on the TIMSAC-78 subroutine NRASPE.

The subroutine ARMASE computes the power spectrum of a

- AR-MA process
- AR process
- MA process

Outputs are given at frequencies $I/(2 * H)$ where $I = 0,1,\dots,H$.

In computing this parametric power spectrum estimate use is made of the TIMSAC-78 subroutine FOUGER. This subroutine computes the Fourier Transform by GOERTZEL's method.

2.8. Subroutine ARMASP

This subroutine

- finds peak amplitudes
- plots all amplitudes versus frequencies
- prints peak amplitudes along with their respective periods
- prints the spectrum

Originally, the printer plot of the power spectrum was performed by the TIMSAC-78 subroutine NRASPE. However, this subroutine has been replaced by the subroutine ARSPLT which was used and referred to in [10].

3. DESCRIPTION OF INPUT

The input of the ARSTU program consists of three parts: parameters, title or heading and time series data.

The parameters, used as arguments or as program control, are inputted by answering the questions displayed on the terminal display monitor. The sequence in which these questions are displayed cannot be changed. Without any exception these parameters are of the integer type.

The title or heading for the time series data may consist of any character; numerical, alphanumerical or special.

The time series data can be inputted in two ways: from the terminal keyboard or from a datafile.

3.1. Displayed Questions

ENTER : NPROG	GENERAL PROG. PAR. 0 = STOP 1 = LEAST SQUARES 2 = MAX. LIKELIHOOD 3 = BAYESIAN
ENTER : LU	OUTPUT DEVICE 1 = DISPLAY 6 = PRINTER
ENTER : ISWLS	PROG. PAR. FOR ARMFIT 0 = MAICE MODEL ONLY (OUTPUT SUPPRESSED) 1 = MAICE MODEL ONLY 2 = ALL AR-MODELS UP TO ORDER LAG
ENTER : ISWML	PROG. PAR. FOR ARMLE PRINT OUT CONTROL NON.LIN.OPT. PROCEDURE 0 = NO 1 = YES

ENTER : ISWBA PROG. PAR. FOR ARBAYS
 0 = TO SUPPRESS THE OUTPUTS
 1 = TO PRINT OUT THE OUTPUTS

ENTER : NCHK CHAR. ROOTS OF THE AUTOREGR. OPERATOR
 0 = NO
 1 = YES

ENTER : MT INPUT OF DATA
 0 = CONSOLE
 1 = FILE

ENTER : N DATA LENGTH (N.LE.300)

ENTER : LAG UPPER LIMIT OF AR-MODEL (LAG.LE.30)

ENTER : NSPAN PROG. PAR. FOR ARMASE
 0 = ARMASE SUPPRESSED
 1 = SPECTRUM (NBRL GIVEN)
 2 = SPECTRAL DENSITIES (UNIT GIVEN)

ENTER : NBRL NUMBER OF LINES SPECTRUM PLOT (NBRL.LE.50)
 UNIT = PXX (MAX)/FLOAT (NBRL)

ENTER : NUNIT UNIT ON Y-AXIS DENSITY PLOT (NUNIT.LE.50)
 UNIT = FLOAT (NUNIT)/1000

ENTER : TITLE OF DATA

ENTER : DATA IN FREE FIELD FORMAT

ENTER : NAME OF DATA FILE

3.2. Conditions, Restrictions and Comments

ISWLS - Will only be displayed if NPROG = 1
 - If ISWLS = 0 the model procedure ARMFIT will produce no
 output at all ! This situation only makes sense if
 NSPAN \neq 0 in which case the spectrum or the spectral
 densities, based on the model fitted by ARMFIT, will be
 given.

- ISWML - Will only be displayed if NPRG = 2
- ISWBA - Will only be displayed if NPRG = 3
 - If ISWBA = 0 the model procedure ARBAYS will produce no output at all ! This situation only makes sense if NSPAN \neq 0, in which case the spectrum or the spectral densities, based on the model fitted by ARBAYS, will be given.
- MT - If MT = 0 the time series data is inputted via the terminal keyboard in FREE FIELD FORMAT.
 - If MT = 1 the time series data is inputted from a datafile on disk.
- N - Restriction: $N \leq 300$
- LAG - Restriction: $LAG \leq 30$
- NSPAN - If NSPAN = 1 the spectrum is computed. The number of lines used for the graphical representation of the spectrum is given by the parameter NBRL.
 The unit on the Y-axis is then given by

$$UNIT = PXX (MAX)/FLOAT (NBRL)$$
 where PXX (MAX) stands for the maximum of the spectrum and NBRL is the number of lines.
 - If NSPAN = 2 the spectral densities are computed. The unit on the Y-axis is determined by the parameter NUNIT. The unit itself is given by

$$UNIT = FLOAT (NUNIT)/1000.$$
 Since the densities are scaled between zero and one the number of lines used in the graphical representation of the spectral densities is defined as the nearest integer to

$$NBRL = 1/UNIT$$
- NBRL - Will only be displayed if NSPAN = 1
- NUNIT - Will only be displayed if NSPAN = 2
- TITLE - May consist of any character; numerical, alphanumerical or special
 - Restriction: the maximum number of characters is 20.

DATA

- Will only be displayed if MT = 0.
- Time series data is inputted via the terminal keyboard in FREE FIELD FORMAT. The input must follow immediately after the message

ENTER: DATA IN FREE FIELD FORMAT

has been displayed on the screen. Each observation must be separated from the other by at least one blank.

NAME

- Will only be displayed if MT = 1.
- Time series data are to be read from a datafile on disk. The name of this datafile must be entered immediately after the message

ENTER: NAME OF DATAFILE

is displayed on the screen (see Appendix I for details about the preparation of this datafile).

4. DESCRIPTION OF OUTPUT

Each output starts with an echo of the input parameters. The sequence in which these parameters are given is: NPROG, ISWML, ISWBA, NCHK, MT, N, LAG, NSPAN, NBRL, NUNIT.

4.1. Output Minimum AIC Procedure (NPROG = 1)

The output for the Least Squares procedure consists of:

- mean and variance of the inputted time series
- order, innovations variance, AIC(M) and AIC(M) - AICMIN for all models up to order LAG
- AR-coefficients and innovation variance for all models up to order LAG
- coefficients and innovation variance for that model for which AIC(M) was minimum
- characteristic roots of the AR-operator of the MAIC model
- spectrum or spectral densities based on the coefficients and innovation variance of the MAIC model
- amplitude, frequency and period for the peak frequencies of the spectrum or spectral densities.

In many situations it might be advantageous to suppress most of the output. This can be done readily by means of the parameters ISWLS, NCHK and NSPAN.

See Section 5 for an example of the output.

4.2. Output Bayesian Procedure (NPROG = 3)

The output for the Bayesian procedure consists of:

- mean and variance of the inputted time series
- order, innovation variance, AIC(M) and AIC(M) - AICMIN for all models up to order LAG
- partial correlation coefficients, weights and integrated weights
- partial correlation coefficients for the Bayesian model, innovation variance, equivalent number of parameters and equivalent AIC
- coefficients and innovation variance for the Bayesian model
- characteristic roots of the AR-operator of the Bayesian model

- spectrum or spectral densities based on the coefficients and innovation variance of the Bayesian model
- amplitude, frequency and period for the peak frequencies of the spectrum or spectral densities.

In many situations it might be advantageous to suppress most of the output. This can be done readily by means of the parameters ISWBA, NCHK and NSPAN.

See Section 6 for an example of the output.

4.3. Output Maximum Likelihood (NPROG = 2)

Not available yet.

5. OUTPUT EXAMPLE : ARMFIT

Input:

NPROG	=	1
ISWLS	=	2
NCHK	=	1
MT	=	1
N	=	156
LAG	=	24
NSPAN	=	1
NBRL	=	25
TITLE	=	&USTRB
NAME	=	&USTRB

Resulting output: See pp. 17-20.

1 2 0 0 1 1 156 24 1 25 0

PROGRAM TIKSAC 78.1.1

AUTOREGRESSIVE MODEL FITTING (SCALAR CASE) ; LEAST SQUARES METHOD BY HOUSEHOLDER TRANSFORMATION

(AUTOREGRESSIVE MODEL)

$$Z(I) = A(1)*Z(I-1) + A(2)*Z(I-2) + \dots + A(M)*Z(I-M) + E(I)$$

WHERE

M: ORDER OF THE MODEL
 E(I): GAUSSIAN WHITE NOISE WITH MEAN 0 AND VARIANCE SD(M).

FITTING UP TO THE ORDER K = 24 IS TRIED
 ORIGINAL DATA INPUT DEVICE MT = 1

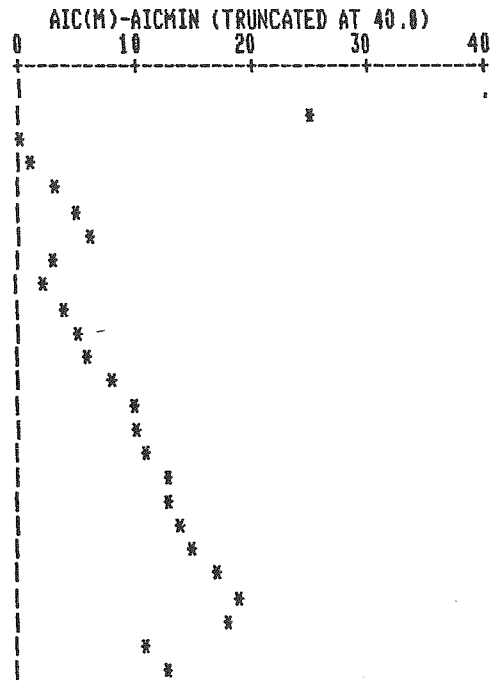
** ORIGINAL DATA **

TITLE -- "AUSTRB

MEAN = .34223525D+01 VARIANCE = .10679650D+01

 MAICE PROCEDURE

ORDER M	INNOVATION VARIANCE SD(M)	AIC(M)	AIC(M)-AICMIN
0	.1196941236D+01	25.730	425.085
1	.5677142792D-01	-374.671	24.684
2	.4638072046D-01	-399.355	0.000
3	.4593079726D-01	-398.642	.713
4	.4592973596D-01	-396.645	2.710
5	.4592916546D-01	-394.646	4.709
6	.4561311683D-01	-393.558	5.797
7	.4402794058D-01	-396.227	3.128
8	.4294280743D-01	-397.521	1.834
9	.4284984885D-01	-395.807	3.548
10	.4261439242D-01	-394.534	4.821
11	.4233428370D-01	-393.405	5.950
12	.4232824520D-01	-391.424	7.931
13	.4227404960D-01	-389.593	9.762
14	.4179959059D-01	-389.083	10.272
15	.4143058217D-01	-388.253	11.102
16	.4132143808D-01	-386.601	12.754
17	.4085919079D-01	-386.086	13.269
18	.4058271636D-01	-384.983	14.373
19	.4014472994D-01	-384.415	14.940
20	.4014415803D-01	-382.417	16.938
21	.4013943480D-01	-380.432	18.923
22	.3929409100D-01	-381.242	18.113
23	.3673279400D-01	-388.139	11.216
24	.3671907067D-01	-386.189	13.166



***** MINIMUM AIC = -.3993550266D+03 ATTAINED AT M = 2 SD(M) = .4638072046D-01 *****


```

..... M A I C E .....
:
:   &USTRB
:
:   COEFFICIENTS          INNOVATION VARIANCE
:   I      A(I)          SD =
:   1      1.4166665153
:   2      -.4380107479
:
:   THIS MODEL WAS FITTED TO THE DATA ( Z( 25), ... ,Z( 156) )
:
.....

```

CHARACTERISTIC ROOTS OF AR OPERATOR $1-A(1)*S- \dots -A(N)*S^*N=0$

REAL	IMAGINARY	SQRT(R**2+I**2)	ARCTAN(I/R)	DEGREE
.45589D+00	.00000D+00	.45589D+00	.00000D+00	0.000
.96077D+00	.00000D+00	.96077D+00	.00000D+00	0.000

6. OUTPUT EXAMPLE : ARBAYS

Input:

```
NPROG = 3
ISWBA = 1
NCHK  = 1
MT     = 1
N      = 156
LAG    = 24
NSPAN  = 1
NBRL   = 25
TITLE  = &USTRB
NAME   = &USTRB
```

Resulting output: See pp. 22-26.

3 0 0 1 1 1 156 24 1 25 0

PROGRAM TIMSAC 78.1.2

EXPONENTIALLY WEIGHTED BAYESIAN AUTOREGRESSIVE MODEL FITTING; SCALAR CASE

(AUTOREGRESSIVE MODEL)

$$Z(I) = A(1)*Z(I-1) + A(2)*Z(I-2) + \dots + A(M)*Z(I-M) + E(I)$$

WHERE

M: ORDER OF THE MODEL
 E(I): GAUSSIAN WHITE NOISE WITH MEAN 0 AND VARIANCE SD(M).

24-TH ORDER BAYESIAN MODEL IS FITTED
 ORIGINAL DATA INPUT DEVICE MT = 1

** ORIGINAL DATA **

TITLE -- "&USTRB

MEAN = .34223525D+01 VARIANCE = .10679650D+01

 BAYESIAN PROCEDURE

ORDER M	INNOVATION SD(M)	VARIANCE	AIC(M)	AIC(M)-AICMIN	AIC(M)-AICMIN (TRUNCATED AT 40.0)
0	.1196941236D+01		25.730	425.085	
1	.5677142792D-01		-374.671	24.684	
2	.4638072046D-01		-399.355	0.000	*
3	.4593079726D-01		-398.642	.713	*
4	.4592973596D-01		-396.645	2.710	*
5	.4592916546D-01		-394.646	4.709	*
6	.4561311683D-01		-393.558	5.797	*
7	.4402794058D-01		-396.227	3.128	*
8	.4294280743D-01		-397.521	1.834	*
9	.4284984885D-01		-395.807	3.548	*
10	.4261439242D-01		-394.534	4.821	*
11	.4233428370D-01		-393.405	5.950	*
12	.4232824520D-01		-391.424	7.931	*
13	.4227404960D-01		-389.593	9.762	*
14	.4179959059D-01		-389.083	10.272	*
15	.4143058217D-01		-388.253	11.102	*
16	.4132143808D-01		-386.601	12.754	*
17	.4085919079D-01		-386.086	13.269	*
18	.4058271636D-01		-384.983	14.373	*
19	.4014472994D-01		-384.415	14.940	*
20	.4014415803D-01		-382.417	16.938	*
21	.4013943480D-01		-380.432	18.923	*
22	.3929409100D-01		-381.242	18.113	*
23	.3673279400D-01		-388.139	11.216	*
24	.3671907067D-01		-386.189	13.166	*

***** MINIMUM AIC = -.3993550266D+03 ATTAINED AT M = 2 SD(M) = .4638072046D-01 *****

< BAYESIAN MODEL >

PARCOR'S

1	.97599671
2	-.42781526
3	-.05068102
4	.00124840
5	.00065056
6	.01340038
7	.02797764
8	.01758022
9	-.00220098
10	-.00167528
11	-.00086858
12	.00005397
13	-.00008572
14	.00017016
15	.00009727
16	-.00003499
17	.00005671
18	.00003535
19	.00003862
20	-.00000125
21	.00000342
22	.00004507
23	.00007741
24	-.00000156

INNOVATION VARIANCE	=	.45516119D-01
EQUIVALENT NUMBER OF PARAMETERS	=	3.430
EQUIVALENT AIC	=	-400.978

BAYESIAN MODEL

&USTRB

COEFFICIENTS

INNOVATION VARIANCE

SD = .4551611868D-01

I	A(I)
1	1.3710814565
2	-.3557354506
3	-.0514885340
4	.0064151945
5	-.0068584541
6	-.0188306624
7	.0029982404
8	.0199598395
9	-.0002076768
10	-.0004715297
11	-.0009649601
12	.0002383221
13	-.0002856324
14	.0000263787
15	.0001653759
16	-.0000974985
17	.0000236567
18	-.0000176558
19	.0000433645
20	.0000140874
21	-.0000309162
22	-.0000616174
23	.0000795461
24	-.0000015576

THIS MODEL WAS FITTED TO THE DATA (Z(25), ... ,Z(156))

CHARACTERISTIC ROOTS OF AR OPERATOR $1-A(1)*S- \dots -A(M)*S^M=0$

REAL	IMAGINARY	SQRT(R**2+I**2)	ARCTAN(I/R)	DEGREE
.19890D-01	.00000D+00	.19890D-01	.00000D+00	0.000
.94010D+00	.00000D+00	.94010D+00	.00000D+00	0.000
-.63213D+00	.88178D-01	.63825D+00	.30030D+01	172.059
-.63213D+00	-.88178D-01	.63825D+00	-.30030D+01	-172.059
.66746D+00	-.11703D+00	.67765D+00	-.17357D+00	-9.945
.66746D+00	.11703D+00	.67765D+00	.17357D+00	9.945
.52090D+00	-.39015D+00	.65081D+00	-.64286D+00	-36.833
.52090D+00	.39015D+00	.65081D+00	.64286D+00	36.833
-.56954D+00	-.26493D+00	.62815D+00	-.27062D+01	-155.054
-.56954D+00	.26493D+00	.62815D+00	.27062D+01	155.054
.22708D-01	-.65185D+00	.65225D+00	-.15360D+01	-88.005
.22708D-01	.65185D+00	.65225D+00	.15360D+01	88.005
.65377D+00	-.28631D+00	.71372D+00	-.41278D+00	-23.651
.65377D+00	.28631D+00	.71372D+00	.41278D+00	23.651
-.48304D+00	-.41878D+00	.63931D+00	-.24273D+01	-139.076
-.48304D+00	.41878D+00	.63931D+00	.24273D+01	139.076
.35430D+00	-.53119D+00	.63850D+00	-.98257D+00	-56.297
.35430D+00	.53119D+00	.63850D+00	.98257D+00	56.297
-.34969D+00	-.53851D+00	.64209D+00	-.21467D+01	-122.998
-.34969D+00	.53851D+00	.64209D+00	.21467D+01	122.998
.19578D+00	-.63906D+00	.66837D+00	-.12735D+01	-72.967
.19578D+00	.63906D+00	.66837D+00	.12735D+01	72.967
-.17497D+00	-.60831D+00	.63297D+00	-.18509D+01	-106.047
-.17497D+00	.60831D+00	.63297D+00	.18509D+01	106.047

7 . OUTPUT EXAMPLE : ARMLE

Not available yet.

APPENDIX I - INPUT DATAFILE

1. The datafile is entered, CReated or REplaced by the EDITR. Each line (record), consisting of 70 characters, must contain 7 observations. Each observation, entered in free format, is separated from the other by at least one blank.

For an example see the input datafile EXF1 on p.29.

2. If the input datafile is entered such that
 - each line (record) is equally divided into 8 fields, each field consisting of 10 characters,
 - the first 7 of these fields are used for time series observations,
 - each observation is separated from the other by at least one blank,
 - the 8th field, i.e. the last 10 characters of the record, is used for the Identification and Sequencing of the data record

this same input datafile can also be used for the MULTISTOCH and MULTITRAN programs.

For an example see the input datafile EXF2 on p.29.

3. Apart from the Identification and the Sequencing the input datafile, resulting from a TSERS' PUNC command, is MULTISTOCH and MULTITRAN compatible. The FORMAT used is (7 (G 9.5, 1X)).

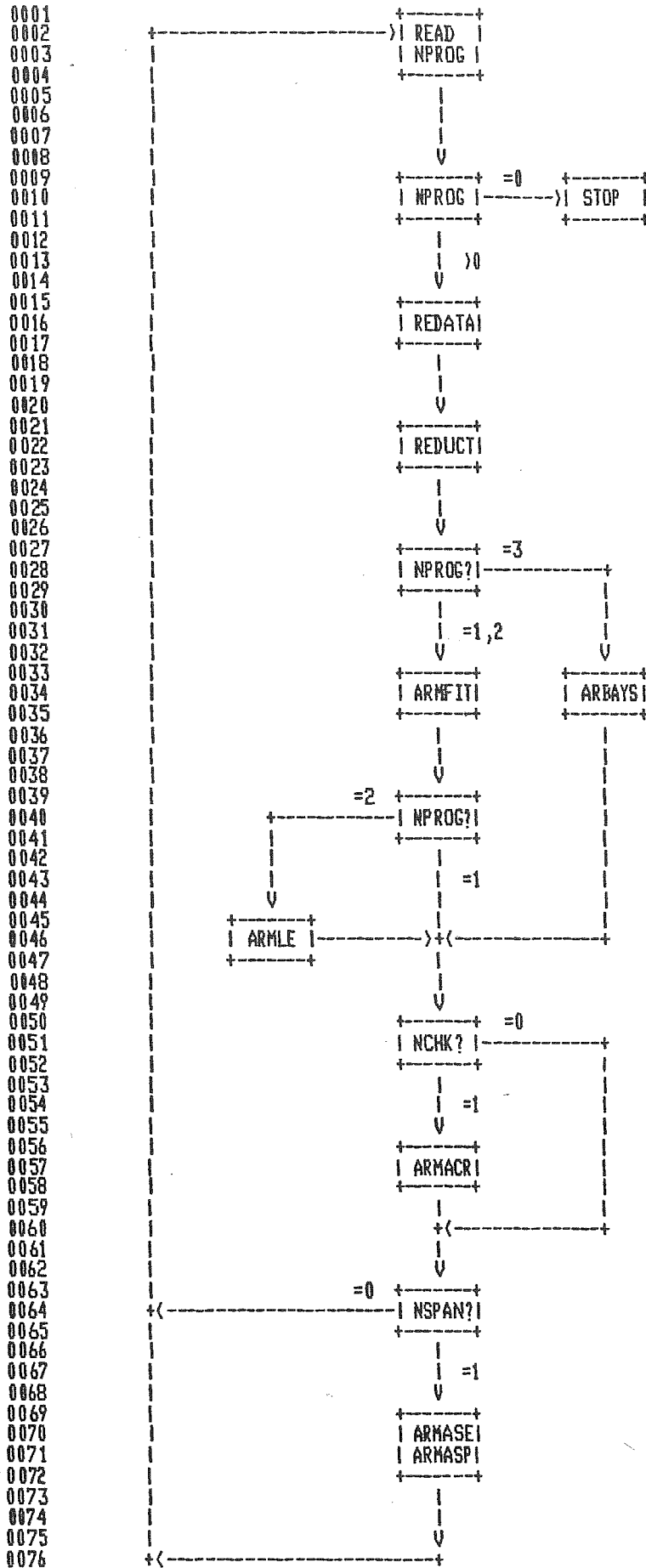
EXF1 T=00004 IS ON CR00007 USING 00003 BLKS R=0000

0001	1	2	3	4	5	6	7	8
0002	1234567890	1234567890	1234567890	1234567890	1234567890	1234567890	1234567890	1234567890
0003								
0004	2.456	2.372	2.310	2.613	2.650	2.527	2.334	
0005	2.606	2.850	2.961	3.000	3.230	3.210	3.165	
0006	3.140	3.113	3.042	3.316	3.165	3.404	3.578	
0007	3.591	3.337	3.102	2.598	1.562	1.354	1.126	
0008	1.046	0.881	0.962	1.686	2.484	2.793	2.756	

EXF2 T=00004 IS ON CR00007 USING 00004 BLKS R=0000

0001	1	2	3	4	5	6	7	8
0002	1234567890	1234567890	1234567890	1234567890	1234567890	1234567890	1234567890	1234567890
0003								
0004	2.456	2.372	2.310	2.613	2.650	2.527	2.334	USTRB 01
0005	2.606	2.850	2.961	3.000	3.230	3.210	3.165	USTRB 02
0006	3.140	3.113	3.042	3.316	3.165	3.404	3.578	USTRB 03
0007	3.591	3.337	3.102	2.598	1.562	1.354	1.126	USTRB 04
0008	1.046	0.881	0.962	1.686	2.484	2.793	2.756	USTRB 05

FLOW T=00004 IS ON CR00007 USING 00016 BLKS R=0000



APPENDIX III : SUBROUTINE MAP

1. REDATA

AXAGET	
	CODE
	OPENF
	READF

2. REDUCT

HUSHLD
SETX1

3. ARMFIT

COMAIC
MAICE
PRINTA
RECOEF

4. ARBAYS

ARCOEF
BAYSPC
BAYSWT
COMAIC
MAICE
PRINTA
SDCOMP

5. ARMLE

DAVIDN			
	FUNCT		
		ARCOEF	
		PARCOR	
		SUBDET	
	HESSIAN		
		INVDET	
	LINEAR		
		FUNCT	
			ARCOEF
			PARCOR
			SUBDET
PRINTA			

6. ARMACR

POLYRT

7. ARMASE

FOUGER

8. ARMASP

PEAKS
PLOTA
PRNT

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