

NONPARAMETRIC TESTS FOR TRENDS IN WATER-QUALITY DATA USING THE STATISTICAL ANALYSIS SYSTEM

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U.S. GEOLOGICAL SURVEY
Open-File Report 83-550

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ABSTRACT

Two nonparametric procedures to test for trends in water-quality data (SEASKEN AND SEASRS) have been developed for the Statistical Analysis System* (SAS). The procedure SEASKEN tests for a monotonic trend in time by a modified form of Kendall's tau, the Seasonal Kendall test. The procedure SEASRS tests for a step trend between two different periods in a time series using a modified form of the Wilcoxon (Mann-Whitney) rank sum test, the Mann-Whitney-Wilcoxon rank sum test for seasonal data. Examples are presented using the two procedures. The source code and user's guide for each of the two procedures are also presented.

Procedures for flow adjusting water-quality data by the SAS procedures REG and SYSREG and techniques for plotting water-quality data as a time series by the SAS procedure PLOT are presented.

Additionally, examples are presented to demonstrate the use of the U.S. Geological Survey procedures QWRETR, DVRETR, QWSAS, and DVINPUT to retrieve data from the Geological Survey WATSTORE system and make it available to SAS.

INTRODUCTION

Increased public concern over the quality of the Nation's rivers in the past several decades has led Federal, State, and local officials to implement or greatly expand existing water-quality monitoring programs. Examples of such programs are the Geological Survey's Benchmark and NASQAN networks (see Briggs, 1978). Most of these monitoring programs have as a goal the detection of trends in water quality.

* Use of brand and firm trade names in this report is for identification purposes only and does not constitute endorsement by the U.S. Geological Survey.

Techniques of trend detection in water-quality data have correspondingly received much attention recently in the literature (see, for example, Fuller and Tsokos, 1971; Lettenmaier, 1976; and Hirsch and others, 1982). Most of the trend detection methods presented to date are based on classical (parametric) hypothesis testing. However, because of the nature of water-quality data (typically skewed, serially correlated, and showing seasonality), many of the assumptions underlying classical hypothesis tests are not met, rendering them inappropriate. (For a discussion of the problems associated with these tests, the reader is referred to Smith and others, 1982.) Recently, however, thinking has begun to shift toward distribution-free (nonparametric) tests that have less restrictive assumptions than their classical counterparts and are therefore less sensitive to the distribution of the water-quality time series.

This report describes procedures in the Statistical Analysis System (SAS) that can appropriately be used to detect trends in water-quality data. The report describes the use of both the standard procedures provided with SAS and two additional procedures, SEASKEN and SEASRS. A working knowledge of SAS by the reader is assumed. The report was written primarily for user's of the U.S. Geological Survey's WATSTORE data base and Amdahl computer system; however, the source code for the SAS macros and procedures are included in the appendices. The user's guide for the SEASKEN and SEASRS procedures are also included in the appendices.

RETRIEVING DATA FROM THE WATSTORE DAILY VALUES AND WATER QUALITY FILES

In order to use SAS on water-quality data, it is first necessary to get the data into a SAS data set. A SAS data set is a collection of data observations addressable by SAS procedures. Each observation has one or more variables associated with it. For more information about SAS data sets, see SAS Institute, Inc.

(1979) or SAS Institute, Inc. (1982a). Data can be entered into a SAS data set from data cards in the job stream or by reading data stored on disk or tape files. When using data from the WATSTORE file, one of the standard Survey retrieval procedures must first be used. One of two Survey applications programs (PROC QWSAS or the DVINPUT macro) is then called to convert the standard retrieval output into a SAS data set. PROC QWSAS is a SAS procedure that produces a SAS data set from the standard water quality file retrieval procedure QWRETR. PROC QWSAS is described in detail in the WATSTORE user's guide, volume 3, chapter IV, section R. DVINPUT is a SAS macro that converts output from the WATSTORE daily values file retrieval, DVRETR, to a SAS data set. Use of the DVINPUT macro is described in the WATSTORE message SAS documentation section (member WRD06). The standard retrieval procedures, DVRETR and QWRETR, are discussed in the WATSTORE user's guide, volumes 1 and 3, respectively.

Figure 1 shows example input using the QWRETR and QWSAS procedures. This example retrieves twelve parameters - two streamflow parameters (00060 & 00061) and ten water-quality constituents - for three stations from the water quality file. Both daily mean streamflow (parameter code 00060) and instantaneous streamflow (00061) should be retrieved. For purposes of this test, the mean streamflow may be used if the instantaneous streamflow is missing. PROC QWSAS creates a SAS data set named DATA1 containing the station identification number and the eleven parameters requested in the QWRETR procedure. The parameter values are stored in variables named Pnnnnn where nnnnn is the WATSTORE parameter codes given in the retrieval list. In addition, the variables YEAR, MONTH, DAY, DATE, DECTIME, and SNAME were requested in the QWSAS statement. The variable DATE is the day on which the sample was collected, in days since January 1, 1960. DECTIME is a decimal number representing the time the sample was collected, in

```

1 //F1 JOB
2 // CLASS
3 /*SETUP 118545/H
4 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
5 // EXEC QWRETR,VOL1=118545
6 //HDR=SYSIN DD *
7 M3 1968100119810930
8 XQWSASX
9 R00060000610041000630000665009150092500930000940009450095070300
10 000060000610041000630000665009150092500930000940009450095070300
11 D 03276500
12 D 03374100
13 D 03378500
14 /*
15 // EXEC WRDSAS
16 //TREND DD DSN=USERID.FILENAME,DISP=OLD
17 //SYSIN DD *
18 PROC QWSAS YEAR MONTH DAY DATE DECTIME SNAME;
19 DATA;SET;IF P00061=, THEN P00061=P00060;DROP P00060;
20 PROC SORT;BY STATION YEAR MONTH DAY;
21 PROC PRINT;BY STATION;
22 VAR YEAR MONTH DAY DECTIME DATE P00061 P00410 P00630 P00665 P00915 P00925 P009
23 30 P00940 P00945 P00950 P70300;
24 DATA TREND;MONTHLY;SET;
25 /*
26 //

```

Figure 1.--Example input for WATSTORE retrieval of water-quality data by the QWRETR and QWSAS procedures.

years. These two forms of sample collection time are useful in plotting the data as time series. Additionally, the variable DATE may be formatted in several different ways (see SAS Institute, Inc., 1982a, p. 409). SNAME is the variable containing the station name.

The variable DECTIME is required to use the SAS procedures SEASKEN and SEASRS. For data sets that do not already include the variable DECTIME, it can be easily generated using the following statement in a SAS data step:

```
DECTIME = YEAR(DT)+(JULDATE(DT)-YEAR(DT)*1000)/365;      (1)
```

where DT is the date the observation was made.

An example use of this is shown in figure 2.

The example in figure 1 also sorts and prints the information retrieved by station. Finally, the data is stored as file TREND.MONTHLY in the data set USERID.FILENAME on a direct access device for later analysis. Line 16 describes the existing file to be used for data storage, and line 24 copies the data to the disk file. For information on creating disk data sets on the USGS Amdahl computer system, see the USGS Computer Users Manual, Chapter 5.

To adapt this example input to a specific application, the user will need to (1) substitute the six digit volume number of the appropriate water quality back file tape for 118545 in lines 3 and 5; (2) substitute the appropriate retrieval dates on the master control card on line 7; (3) change the retrieval and output list as desired on lines 9 and 10; (4) substitute the desired station numbers in lines 11 through 13 (and delete or add D cards as required); (5) substitute an appropriate data set name in the DSN = field on line 16; and (6) change the variables list in lines 22 and 23 to agree with the parameters listed in the retrieval list and PROC QWSAS optional variables.

```

1 //F2 JOB
2 // CLASS
3 /*SETUP 115620/H
4 //PROCLIB DD DSN=WRD,PROCLIB,DISP=SHR
5 // EXEC DVRETR,AGENCY=USGS,VOL1=115620
6 //HDR.SYSIN DD *
7 M3 1970100119800930
8 R000608015480155
9 F00003
10 D 03365500
11 /*
12 // EXEC WRDSAS,MACRO=DV,DSN='&&BKREC',DSN1=NULLFILE,DSN2=NULLFILE
13 //TREND DD DSN=USERID.FILENAME,DISP=OLD
14 //SYSIN DD *
15 DVINPUT _SNAME
16 DATA DATAA(RENAME=(VALUE=P00060)) DATAB(RENAME=(VALUE=P80154)) DATAC(RENAME=(VAL
17 UE=P80155));SET;
18 IF PARMCODE=60 THEN OUTPUT DATAA;
19 IF PARMCODE=80154 THEN OUTPUT DATAB;
20 IF PARMCODE=80155 THEN OUTPUT DATAC;
21 DROP PARMCODE;
22 PROC SORT DATA=DATAA;BY STATION DATE;
23 PROC SORT DATA=DATAB;BY STATION DATE;
24 PROC SORT DATA=DATAC;BY STATION DATE;
25 DATA TEMP;MERGE DATAA DATAB DATAC;BY STATION DATE;
26 FORMAT DATE YMMDD8.;
27 J=JULDATE(DATE);Y=YEAR(DATE);
28 DECTIME=Y+(J-Y*100)/365;
29 LABEL P80154=SUSPENDED SEDIMENT CONCENTRATION (MG/L)
30 P80155=SUSPENDED SEDIMENT DISCHARGE (TONS/DAY)
31 P00060=STREAMFLOW (CFS);
32 PROC SORT;BY STATION DATE;
33 PROC PRINT;BY STATION;VAR DATE DECTIME P00060 P80154 P80155;
34 DATA TREND.DAILY;SET TEMP;
35 /*
36 //

```

Figure 2.--Example input for WATSTORE retrieval of streamflow data by the DVRETR and DVINPUT procedures.

The lines in the QWRETR example in figure 1 do the following:

- Line 1. Job control.
2. Class (extension of line 1).
3. Instructs the computer operator to mount the specified backfile tape.
4. Invokes the WRD cataloged procedure library.
5. Invokes the procedure QWRETR and specifies the mounted backfile tape to be used in the retrieval.
6. Establishes that data for QWRETR follow.
7. Specifies that data from both the current and backfile are to be retrieved for the period October 1, 1968, to September 30, 1981.
8. Specifies that QWSAS will be invoked as an application program.
9. Restricts the retrieval to the listed parameters only.
10. Restricts the output list to the listed parameters only.
- 11-13. Specifies the stations to be included in the retrieval.
14. Job control language step separator card (step delimiter).
15. Invokes the procedure WRDSAS (WRD modified version of SAS).
16. Defines existing direct access data set to be used for storage of the retrieved data.
17. Establishes that SAS instructions follow.
18. Invokes the procedure QWSAS and requests the variables YEAR, MONTH, DAY, DATE, DECTIME, and SNAME to be included in the data set created in line 23 below.
19. Uses mean streamflow (parameter code 00060) if instantaneous streamflow (00061) is missing, and discards the mean streamflow parameter.
20. Sorts data set in order of variables listed in BY statement.
21. Prints data set.

- 22-23. Selects and orders variables to be included in print of data set.
24. Creates file TREND.MONTHLY containing the retrieved data and stores it in the data set USERID.FILENAME.
25. Step delimiter.
26. End of job.

Figure 2 shows example input using the DVRETR and DVINPUT macro procedures. This example retrieves streamflow, suspended-sediment concentration, and suspended-sediment discharge from the daily values file for one station. The DVINPUT macro creates a SAS data set containing the station identification number, parameter code, value, date, and the optional variable requested on the DVINPUT statement, station name. The variable PARMCODE contains the values of the WATSTORE parameter codes in the retrieval list. This format of the data set is awkward since all the variables requested in the retrieval list are combined into different observations of the variable PARMCODE. The statements in lines 16 through 31 convert the initial SAS data set into one containing the station identification number, station name, date, P00060, P80154, and P80155. In addition, lines 27 and 28 add the variables Y (year), J (Julian date), and DECTIME for each observation, and lines 29 through 31 add variable labels for suspended-sediment concentration, suspended-sediment discharge, and streamflow.

This example also sorts and prints the data contained in the modified data set named TEMP. Finally, the data set is stored as the file TREND.DAILY on the data set USERID.FILENAME on a direct access device.

Note that DVINPUT is a SAS macro and not a SAS procedure. It is not preceded by the statement PROC or followed by a semicolon.

To adapt this example setup to a specific application, the user will need to 1) substitute the six-digit volume number of the appropriate daily values backfile tape for 115620 in lines 3 and 5; 2) substitute the appropriate retrieval dates on the master control card on line 7; 3) change the retrieval and output list as desired on lines 8 and 9; 4) substitute the desired station numbers in line 10 (add or delete D cards as required); 5) change lines 18-20 and 29-31 appropriately; 6) substitute an appropriate data set name in the DSN = field on line 13; and 7) change the variable list in line 33 to agree with the parameters listed in the retrieval list and the DVINPUT optional variables.

The lines in the DVRETR example input of figure 2 do the following:

- Line 1. Job control.
2. Class (extension of line 1).
3. Instructs the computer operator to mount the specified backfile tape.
4. Invokes the WRD cataloged procedure library.
5. Invokes the procedure DVRETR with AGENCY=USGS being the agency code and specifies the mounted backfile tape to be used in the retrieval.
6. Establishes that data cards for DVRETR follow.
7. Establishes that data from both the current and backfile are to be retrieved for the period October 1, 1970, to September 30, 1980.
8. Restricts the retrieval to the listed parameters only.
9. Restricts retrieval to listed statistics codes only.
10. Specifies the station to be included in the retrieval.
11. Step delimiter.
12. Invokes the procedure WRDSAS and defines the temporary data set containing retrieval data and supplies the macro DV.

13. Defines existing direct access data set to be used for storage of the retrieved data.
14. Establishes that SAS instructions follow.
15. Invokes the macro DVINPUT and requests that the variable SNAME be included in the created SAS data set.
- 16-31. Converts the SAS data set.
32. Sorts data set in order of variables listed in BY statement.
33. Prints the variables in the VAR statement in the order they are listed in the VAR statement.
34. Creates file TREND.DAILY containing the retrieved data and stores it in the data set USERID.FILENAME.
35. Step delimiter.
36. End of job.

DETERMINING THE RELATIONSHIP BETWEEN WATER-QUALITY CONSTITUENTS AND STREAMFLOW

Quite frequently, concentrations of water-quality constituents are related to streamflow. When a water-quality constituent and streamflow are related, apparent trends in water quality may be due only to fluctuations in streamflow rather than to changes in the processes that affect the introduction and fate of a given constituent in the stream. For example, consider a stream where dissolved solids and streamflow are negatively correlated. That is, as streamflow increases dissolved solids decrease and vice versa. During a period of drought, high dissolved solids concentrations would be expected. If this period of drought was followed by a period of wet weather, a decrease in dissolved solids concentrations would be expected. If such a time series was tested for trend in dissolved solids concentration, a significant downtrend would be

indicated. However, such a trend could be entirely attributable to the fluctuation in streamflow during the period. In order to test for trends in the processes affecting dissolved solids during the period, it would be necessary to remove the effect of streamflow. Flow adjustment is an attempt to remove a major source of variation in water quality (streamflow) which may be masking those variations attributable to changes in the constituent inputs to the stream or in the processes occurring in the stream.

Smith and others (1982) described a flow-adjustment procedure suitable for this purpose. Their approach is to develop a time series of flow-adjusted concentrations (FAC) and to test that series for trend. FAC is defined as the actual concentration (C) minus the expected concentration (\hat{C}) predicted from the discharge (Q) relationship. The FAC should be randomly distributed with a mean of zero over the period of record if no change in the processes that affect the water-quality constituent have occurred. Of course, for some constituents - e.g., biological - adjustment by some other variable - e.g., solar radiation or air temperature - might be appropriate. Where some of the data are reported as "less than" a detection limit, these approaches to flow adjustment are not valid. If only a very few values are reported as "less than," then flow adjustment could be used provided some sensitivity analysis were done to check that the choice of values to use in place of the "less than" was not very influential in the overall results.

Some common models used for flow adjustment include the following:

- (2) $\hat{C} = a + b Q$ linear
- (3) $\hat{C} = a + b \ln(Q)$ log-linear
- (4) $\hat{C} = a + b \frac{1}{1 + \beta Q}$ hyperbolic, β a constant typically in the range $10^{-3} Q^{-1} \leq \beta \leq 10^2 Q^{-1}$, where \bar{Q} is mean discharge
- (5) $\hat{C} = a + b \frac{1}{Q}$ inverse
- (6) $\hat{C} = a + b_1 Q + b_2 Q^2$ quadratic

A good guide to selecting a flow adjustment equation is the R^2 value, but one should check plots of the predicted (\hat{C}) and observed (C) values versus Q (a log Q scale is usually desirable), and plots of the residual versus the predicted (Q) to confirm that the relationship fits well and is not excessively heteroscedastic (e.g., variance increases as predicted concentration increases). For many constituents (particularly suspended constituents or biological ones like bacteria or plankton), these models may be inadequate, because the constituents are very heteroscedastic. In these cases, models based on fitting the log concentrations may be preferred.

Candidate models include the following:

- (7) $\ln \hat{C} = a + b \ln Q$ log-log
 (8) $\ln \hat{C} = a + b_1 \ln Q + b_2 (\ln Q)^2$ log-quadratic log

Deciding between a model based on concentration (equations 2-6) and one based on log concentration (equation 7 or 8) should not be based on R^2 values. Rather, the decision should be based on examination of residuals plots.

If none of the models considered results in a significant fit, as determined from the probability values for the t statistics in the cases with one explanatory variable (equations 2-5 or 7) or the probability value of the F statistic (equations 6 or 8), then no flow adjustment should be performed.

If one of the linear models (equations 2-6) is used, the residuals (Flow Adjusted Concentrations) are defined as $C - \hat{C}$. If one of the log models (equations 7 - 8) is used, the residuals are $\ln C - \ln \hat{C}$. Note that in the former case the residuals have the dimensions of C (typically mg/L), but in the latter case they are dimensionless. This has important implications for the interpretation of trend test results. If log models are used and a slope is estimated in Proc SEASKEN, it must be transformed as follows: If B is the slope

value reported by Proc SEASKEN, then $(e^B-1) \cdot 100$ is the change in percent per year. If log models are used and a step change is being considered and D is the difference (mean FAC of period 2 minus mean FAC of period 1), then the percentage change from period 1 to period 2 would be $(e^D-1) \cdot 100$.

The following section provides two examples of SAS jobs to search for appropriate flow adjustment models. The first, using the macro ABREGMAC, simply estimates the regression equation and computes the usual summary statistics; the second, using the macro REGMAC, provides extensive diagnostics on the results. These macros are intended to be illustrative of the kinds of analyses one may want to run. Knowledge gained by working with a particular data set should dictate variations of these that one may choose to pursue.

The SAS job listing and output in figure 3 follow the procedures used by Smith and others (1982) but includes several other models as well. It considers all of the models defined by equations 2-8. Plots of C versus Q and log C versus log Q are produced to aid in model selection.

Simple regression is used to estimate the coefficients a and b of each model, to calculate R^2 (the fraction of the variance of the dependent variable explained by the given function of Q) and p (the probability of erroneously rejecting the null hypothesis that $b = 0$). β is equal to $10^{(-2.5 - \beta^*)}$ where β^* is the interger part of $\log_{10} \bar{Q}$, where \bar{Q} is the mean streamflow. Eight different hyperbolic models are generated by incrementing the initial value of β by a factor of $10^{0.5}$ seven times.

The macro REGDATA (lines 6-43) calculates the necessary functions of Q needed for the linear regressions. The macro ABREGMAC (lines 44-64) does the linear regressions using the SAS regression procedure SYSREG (SAS Institute, Inc., 1979, p. 403). Lines 65-67 create a data set named DATAA with only sulfate and streamflow data for station 03374100. Line 68 sorts this newly

```

1 //F3 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //TREND DD DSN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 MACRO REGDATA
8 OPTIONS NOMACROGEN;
9 PROC MEANS DATA=FILENAME NOPRINT MEAN;BY STATION;VAR P00061;
10 OUTPUT OUT=MEANQ MEAN=M00061;
11 PROC SORT DATA=MEANQ;BY STATION;
12 DATA INPDATA;MERGE MEANQ FILENAME;BY STATION;
13 *GET LOG OF DEPENDENT VARIABLE;
14 LNDEPVAR=LOG(DEPVAR);
15 *GET LOG OF THE FLOW;
16 L00061=LOG(P00061);
17 *BETA = THE INTEGER PORTION OF THE BASE 10 LOG OF THE MEAN FLOW;
18 BETA=INT(LOG10(M00061));
19 *BN=DIFFERENT VALUES USED IN THE HYPERBOLIC FUNCTIONS;
20 B1=10**(-2.5-BETA);
21 B2=B1*10**0.5;
22 B3=B2*10**0.5;
23 B4=B3*10**0.5;
24 B5=B4*10**0.5;
25 B6=B5*10**0.5;
26 B7=B6*10**0.5;
27 B8=B7*10**0.5;
28 *HYPERUNQ=DIFFERENT HYPERBOLIC FUNCTIONS;
29 HYPERB1Q=1/(1+B1*P00061);
30 HYPERB2Q=1/(1+B2*P00061);
31 HYPERB3Q=1/(1+B3*P00061);
32 HYPERB4Q=1/(1+B4*P00061);
33 HYPERB5Q=1/(1+B5*P00061);
34 HYPERB6Q=1/(1+B6*P00061);
35 HYPERB7Q=1/(1+B7*P00061);
36 HYPERB8Q=1/(1+B8*P00061);
37 *INVQ= THE INVERSE OF THE FLOW;
38 INVQ=1/P00061;
39 *Q2=SQUARE OF THE FLOW;
40 Q2=P00061**2;
41 *LQ2=SQUARE OF THE LOG OF THE FLOW;
42 LQ2=L00061**2;
43 PROC SORT DATA=INPDATA;BY STATION;
44 %
45 MACRO ABREGMAC
46 OPTIONS NOMACROGEN;
47 PROC SYSREG DATA=INPDATA;BY STATION;
48 LTNEAR=MODEL DEPVAR=P00061;
49 LOGLINE=MODEL DEPVAR=L00061;

```

Figure 3.--Example input and output of a SAS procedure to do flow adjustment regressions-abbreviated version.

```

51 HYPER2:MODEL DEPVAR=HYPERB2Q;
52 HYPER3:MODEL DEPVAR=HYPERB3Q;
53 HYPER4:MODEL DEPVAR=HYPERB4Q;
54 HYPER5:MODEL DEPVAR=HYPERB5Q;
55 HYPER6:MODEL DEPVAR=HYPERB6Q;
56 HYPER7:MODEL DEPVAR=HYPERB7Q;
57 HYPER8:MODEL DEPVAR=HYPERB8Q;
58 INVERS:MODEL DEPVAR=INVQ;
59 QUAD: MODEL DEPVAR=P00061 Q2;
60 LOGLOG:MODEL LNDEPVAR=L00061;
61 LOGQAD:MODEL LNDEPVAR=L00061 LQ2;
62 PROC PLOT DATA=INPDATA;BY STATION;
63 PLOT DEPVAR*P00061;
64 PLOT LNDEPVAR*L00061;
65 %
66 DATA DATA;SET TREND.MONTHLY;
67 IF STATION= ' 03374100 ' ;
68 KEEP STATION SNAME P00945 P00061;
69 PROC SORT;BY STATION;
70 MACRO FILENAME DATA%
71 MACRO DEPVAR P00945%
72 MACRO LNDEPVAR L00945%
73 REGDATA
74 ABREGMAC
75 /*
76 //

```

MODEL: L INEAR SSE 9981.852 F RATIO 71.46
 DFE 96 PROB>F 0.0001
 DEP VAR: P00945 MSE 103.977679 R-SQUARE 0.4267
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T RATIO	PROB> T	VARIABLE LABEL
INTERCEPT	1	60.288824	1.418811	42.4925	0.0001	STREAMFLOW, INSTANTANEDUS (CFS)
P00061	1	-0.0005675	0.000671385	-8.4539	0.0001	

MODEL: LOGLIN SSE 7440.682 F RATIO 128.65
 DFE 96 PROB>F 0.0001
 DEP VAR: P00945 MSE 77.507101 R-SQUARE 0.5727
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T RATIO	PROB> T	VARIABLE LABEL
INTERCEPT	1	136.455261	7.495376	18.2053	0.0001	
L00061	1	-9.317384	0.821471	-11.3423	0.0001	

MODEL: HYPER1 SSE 9943.372 F RATIO 72.11
 DFE 96 PROB>F 0.0001
 DEP VAR: P00945 MSE 103.576787 R-SQUARE 0.4289
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T RATIO	PROB> T	VARIABLE LABEL
INTERCEPT	1	-1771.47	214.748414	-8.2491	0.0001	
HYPERB1Q	1	1831.854	215.727945	8.4915	0.0001	

MODEL: HYPER2 SSE 9864.229 F RATIO 73.45
 DFE 96 PROB>F 0.0001
 DEP VAR: P00945 MSE 102.752388 R-SQUARE 0.4335
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T RATIO	PROB> T	VARIABLE LABEL
INTERCEPT	1	-544.097240	69.564112	-7.8215	0.0001	
HYPERB2Q	1	604.667673	70.551800	8.5705	0.0001	

MODEL: HYPER3 SSE 9645.029 F RATIO 77.31
 DFE 96 PROR>F 0.0001
 DEP VAR: P00945 MSE 100.469055 R-SQUARE 0.4461
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROR>|T| VARIABLE LABEL

INTERCEPT 1 -155.241972 23.597084 -6.5789 0.0001
 HYPERB3Q 1 216.378395 24.609842 8.7924 0.0001

MODEL: HYPER4 SSE 9156.534 F RATIO 86.55
 DFE 96 PROR>F 0.0001
 DEP VAR: P00945 MSE 95.380563 R-SQUARE 0.4741
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROR>|T| VARIABLE LABEL

INTERCEPT 1 -30.563720 8.933707 -3.4212 0.0009
 HYPERB4Q 1 93.218221 10.01921 9.3033 0.0001

MODEL: HYPER5 SSE 8437.529 F RATIO 102.11
 DFE 96 PROR>F 0.0001
 DEP VAR: P00945 MSE 87.890932 R-SQUARE 0.5154
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROR>|T| VARIABLE LABEL

INTERCEPT 1 11.803547 4.093053 2.8838 0.0049
 HYPERB5Q 1 54.320665 5.375720 10.1048 0.0001

MODEL: HYPER6 SSE 7884.815 F RATIO 115.99
 DFE 96 PROR>F 0.0001
 DEP VAR: P00945 MSE 82.133486 R-SQUARE 0.5472
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROR>|T| VARIABLE LABEL

INTERCEPT 1 28.670546 2.355140 12.1736 0.0001
 HYPERB6Q 1 44.303562 4.113583 10.7701 0.0001

MODEL: HYPER7 SSE 7889.606 F RATIO 115.87
 DEP VAR: P00945 DFE 96 PR>F 0.0001
 MSE 42.183394 R-SQUARE 0.5469
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROB>|T| VARIABLE LABEL

INTERCEPT 1 36.901791 1.678295 21.9877 0.0001
 HYPERB7Q 1 49.306521 4.580650 10.7641 0.0001

MODEL: HYPERR SSE 8576.029 F RATIO 98.91
 DEP VAR: P00945 DFE 96 PR>F 0.0001
 MSE 89.333636 R-SQUARE 0.5075
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROB>|T| VARIABLE LABEL

INTERCEPT 1 41.562157 1.421870 29.2306 0.0001
 HYPERB8Q 1 72.286630 7.268460 9.9452 0.0001

MODEL: INVERS SSE 10538.97 F RATIO 62.61
 DEP VAR: P00945 DFE 96 PR>F 0.0001
 MSE 109.780918 R-SQUARE 0.3947
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROB>|T| VARIABLE LABEL

INTERCEPT 1 46.117272 1.296412 35.5730 0.0001
 INVQ 1 27804.92 3514.116 7.9124 0.0001

MODEL: QUAD SSE 8825.442 F RATIO 46.21
 DEP VAR: P00945 DFE 95 PR>F 0.0001
 MSE 92.899808 R-SQUARE 0.4931
 SULFATE DISSOLVED (MG/L AS SO4)

VARIABLE DF PARAMETER ESTIMATE STANDARD ERROR T RATIO PROB>|T| VARIABLE LABEL

INTERCEPT 1 64.405530 1.777655 36.2306 0.0001
 P00061 1 -0.00117688 0.000184011 -6.3957 0.0001
 Q2 1 1.06125E-08 3.00799E-09 3.5281 0.0006

STREAMFLOW, INSTANTANEOUS (CFS)

STATION IDENTIFICATION NUMBER=033374100

MODEL: LOGLOG F RATIO 104.24
 DEP VAR: L00945 PROB>F 0.0001
 R-SQUARE 0.5206

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T RATIO	PROB> T	VARIABLE LABEL
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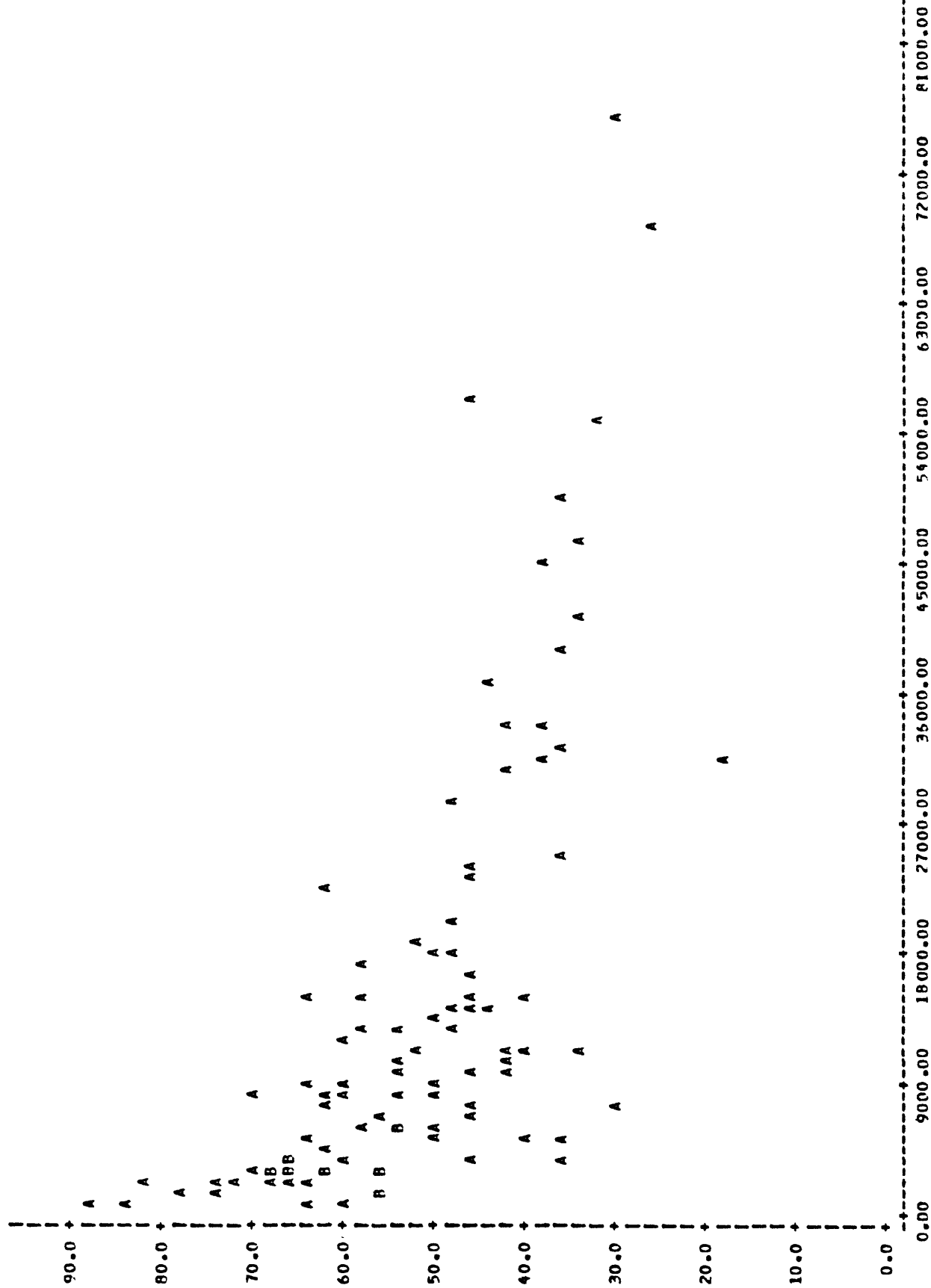
INTERCEPT	1	5.582358	0.164343	33.9676	0.0001	
L00061	1	-0.183890	0.018012	-10.2096	0.0001	

MODEL: LOGQAD F RATIO 52.89
 DEP VAR: L00945 PROB>F 0.0001
 R-SQUARE 0.5268

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T RATIO	PROB> T	VARIABLE LABEL
----------	----	--------------------	----------------	---------	---------	----------------

INTERCEPT	1	4.407203	1.059222	4.1608	0.0001	
L00061	1	0.082749	0.238112	0.3475	0.7290	
L02	1	-0.014901	0.013269	-1.1230	0.2643	

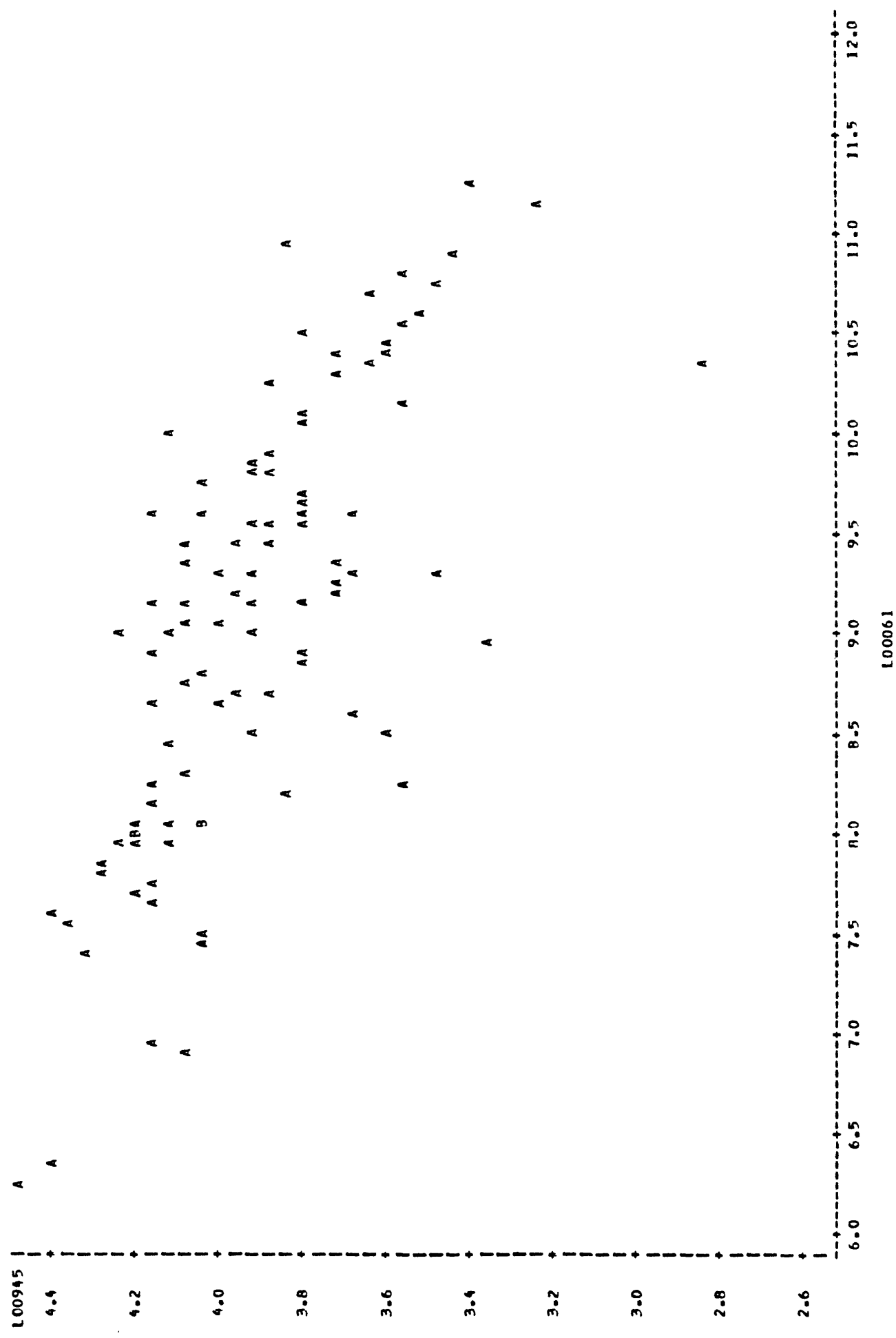
FLOT OF P00945*P00061 LEGEND: A = 1 OBS, B = 2 OBS, ETC.



STREAMFLOW, INSTANTANEOUS (CFS)

NOTE: 5 OBS HAD MISSING VALUES

PLOT OF L00945*L00061 LEGEND: A = 1 OBS, B = 2 OBS, ETC.



NOTE: 5 OBS HAD MISSING VALUES

created data set by station and SNAME (sorting by station and SNAME are required by REGDATA and ABREGMAC). The macros REGDATA and ABREGMAC are written with dummy names for the data set and dependent variables. Lines 69-71 substitute the desired data set and variable names into these macros. In this example, the flow-adjustment procedure will be done on the data set DATAA for the variable P00945 (sulfate). The macro LNDEPVAR inserts a name for the natural logarithm of the dependent variable. The argument of this macro may be any unused variable name the user chooses for the log of the variable given in line 70. The procedures could have been done on any number of variables by repeating the statements in lines 69-73 and making changes in the data set and dependent variables. (For more information on the use of a macro to substitute a variable for dummy names, see SAS Institute, Inc., 1979, p. 12.) The statement in line 72 executes the macro REGDATA and the statement in line 73 executes the macro ABREGMAC.

SYSREG (line 46) outputs for each of the models: (1) the error sum of squares (SSE); (2) the error degrees of freedom (DFE); (3) the error mean square (MSE); (4) the model F ratio and Prob >F (test and significance probability that all parameters except the intercept are zero); (5) the model R-square (R^2); and (6) the degrees of freedom, estimate, standard error, and T ratio and Prob >|T| for parameters in the model. The R-square and Prob >|T| for the function of Q in the model correspond to the R^2 and p value in the Smith and others (1982) flow-adjustment procedure. Plots of streamflow versus sulfate, in both arithmetic and log form, are shown in figure 3.

Figure 4 shows a listing and output of a sample SAS job that does the flow-adjustment procedure using the macro REGMAC (line 44). In addition to providing R^2 and p for each model, the program provides several diagnostic

```

1 //F4 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WKDSAS,DSMT=NULLFILE,DSN2=NULLFILE
5 //TREND DD DSN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 MACRO REGDATA
8 OPTIONS NOMACROGEN;
9 PROC MEANS DATA=FILENAME NOPRINT MEAN;BY STATION;VAR P00061;
10 OUTPUT OUT=MEANQ MEAN=M00061;
11 PROC SORT DATA=MEANQ;BY STATION;
12 DATA INPDATA;MERGE MEANQ FILENAME;BY STATION;
13 *GET LOG OF DEPENDENT VARIABLE;
14 LNDEPVAR=LOG(DEPVAR);
15 *GET LOG OF THE FLOW;
16 L00061=LOG(P00061);
17 *BETA = THE INTEGER PORTION OF THE BASE 10 LOG OF THE MEAN FLOW;
18 BETA=INT(LOG10(M00061));
19 *BN=DIFFERENT VALUES USED IN THE HYPERBOLIC FUNCTIONS;
20 B1=10**(-2.5-BETA);
21 B2=B1*10**0.5;
22 B3=B2*10**0.5;
23 B4=B3*10**0.5;
24 B5=B4*10**0.5;
25 B6=B5*10**0.5;
26 B7=B6*10**0.5;
27 B8=B7*10**0.5;
28 *HYPERBNQ=DIFFERENT HYPERBOLIC FUNCTIONS;
29 HYPERB1Q=1/(1+B1*P00061);
30 HYPERB2Q=1/(1+B2*P00061);
31 HYPERB3Q=1/(1+B3*P00061);
32 HYPERB4Q=1/(1+B4*P00061);
33 HYPERB5Q=1/(1+B5*P00061);
34 HYPERB6Q=1/(1+B6*P00061);
35 HYPERB7Q=1/(1+B7*P00061);
36 HYPERB8Q=1/(1+B8*P00061);
37 *INVQ= THE INVERSE OF THE FLOW;
38 INVQ=1/P00061;
39 *Q2=SQUARE OF THE FLOW;
40 Q2=P00061**2;
41 *LQ2=SQUARE OF THE LOG OF THE FLOW;
42 LQ2=L00061**2;
43 PROC SORT DATA=INPDATA;BY STATION;
44 *
45 MACRO REGMAC
46 OPTIONS NOMACROGEN NOOVP;
47 PROC REG DATA=INPDATA;BY STATION;
48 ID P00061;
49 TITLE OUTPUT FROM MODLABEL: Y=FLOW;
50 MODLABEL:MODEL Y = FLOW / R;

```

Figure 4.--Example input and output of a SAS procedure to do flow adjustment regressions-
extended version.

```

51 OUTPUT OUT=REGDAT1 P=P R=R STUDENT=SR COOKD=COOKD;
52 PROC SORT DATA=REGDAT1 ;BY STATION;
53 PROC PLOT DATA=REGDAT1 ;BY STATION;
54 PLOT P*LU0061='P' Y*LU0061='O' / OVERLAY;
55 PLOT R*P / VREF=0;
56 PROC UNIVARIATE DATA=REGDAT1 PLOT NORMAL;VAR R;
57 %
58 DATA DATA;SET TREND,MONTHLY;
59 IF STATION=' 03374100 ';;
60 KEEP STATION SNAME P00945 P00061;
61 PROC SORT;BY STATION;
62 MACRO FILENAME DATA%
63 MACRO DEPVAR PU0945%
64 MACRO LNDEPVAR LU0945%
65 REGDATA
66 MACRO MODLABEL LINEAR%
67 MACRO Y DEPVAR%
68 MACRO FLOW P00061%
69 REGMAC
70 MACRO MODLABEL LOGLIN%
71 MACRO Y DEPVAR%
72 MACRO FLOW LU0061%
73 REGMAC
74 MACRO MODLABEL LOGLOG%
75 MACRO Y LNDEPVAR%
76 MACRO FLOW LU0061%
77 REGMAC
78 / *
79 //

```

MODEL: LINEAR
DEP VARIABLE: P00945 SULFATE DISSOLVED (MG/L AS SO4)

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB>F
MODEL	1	7429.983	7429.983	71.459	0.0001
ERROR	96	9981.852	103.978		
C TOTAL	97	17411.835			

ROOT MSE 10.196942 R-SQUARE 0.4267
DEP MEAN 52.040815 ADJ R-SQ 0.4207
C.V. 19.59412

VARIABLE	DF	PAFAMETER ESTIMATE	STANDARD ERROR	T FOR HO: PARAMETER=0	PROB > T	VARIABLE LABEL
INTERCEP	1	60.288824	1.418811	42.492	0.0001	INTERCEPT
P00061	1	-0.0005675	0.000671385	-8.453	0.0001	STREAMFLOW, INSTANTANEOUS (CFS)

OBS	ID	ACTUAL	PREDICT VALUE	STD FRR PREDICT	RFSIDUAL	STD FRR RESIDUAL	STUDENT RESIDUAL	COOK'S D
1	12799.97	58.000	53.025	1.037	4.975	10.144	0.490	0.001
2	55899.89	46.000	28.566	2.962	17.434	9.757	1.787	0.147
3	28399.94	49.000	44.172	1.388	4.828	10.102	0.478	0.002
4	9299.98	59.000	55.011	1.088	3.989	10.139	0.393	0.001
5	15299.97	45.000	51.606	1.031	-6.606	10.145	-0.651	0.002
6	31399.93	17.000	42.469	1.531	-25.469	10.081	-2.526	0.074
7	5989.99	53.000	56.890	1.179	-3.890	10.129	-0.384	0.001
8	3029.99	68.000	58.569	1.287	9.431	10.115	0.932	0.007
9	2250.00	67.000	59.012	1.319	7.988	10.111	0.790	0.005
10	2600.00	73.000	58.813	1.305	14.187	10.113	1.403	0.016
11	6360.00	58.000	56.680	1.167	1.320	10.130	0.130	0.000
12	14799.96	57.000	51.890	1.030	5.110	10.145	0.504	0.001
13	22500.00	61.000	47.520	1.161	13.480	10.131	1.331	0.012
14	33599.93	42.000	41.221	1.643	0.779120	10.064	0.077	0.000
15	37000.00	44.000	39.291	1.826	4.709	10.032	0.469	0.004
16	14100.00	49.000	52.287	1.030	-4.287	10.145	-0.423	0.001
17	11700.00	59.000	53.649	1.047	5.351	10.143	0.528	0.001
18	4740.00	61.000	57.599	1.222	3.401	10.123	0.336	0.001
19	7979.98	41.000	55.760	1.120
20	11300.00	45.000	53.876	1.053	-12.876	10.142	-1.270	0.009
21	14299.96	45.000	52.174	1.030	-7.174	10.145	-0.707	0.003
22	23199.95	45.000	47.123	1.183	-2.123	10.128	-0.210	0.000
23	20299.95	48.000	48.769	1.100	-0.768611	10.137	-0.076	0.000
24	38999.92	35.000	38.156	1.939	-3.156	10.011	-0.315	0.002
25	18099.96	50.000	50.017	1.058	-0.017103	10.142	-0.002	0.000
26	45299.91	36.000	34.581	2.308	3.419	9.932	0.344	0.003
27	14899.97	64.000	51.833	1.030	12.167	10.145	1.199	0.007
28	9709.98	53.000	54.778	1.080	-1.778	10.140	-0.175	0.000
29	6549.99	56.000	56.572	1.161	-0.571709	10.131	-0.056	0.000
30	980.00	60.000	59.733	1.374	0.267325	10.104	0.076	0.000

OUTPUT FROM LINEAR: P00945= P00061
STATION IDENTIFICATION NUMBER=03374100

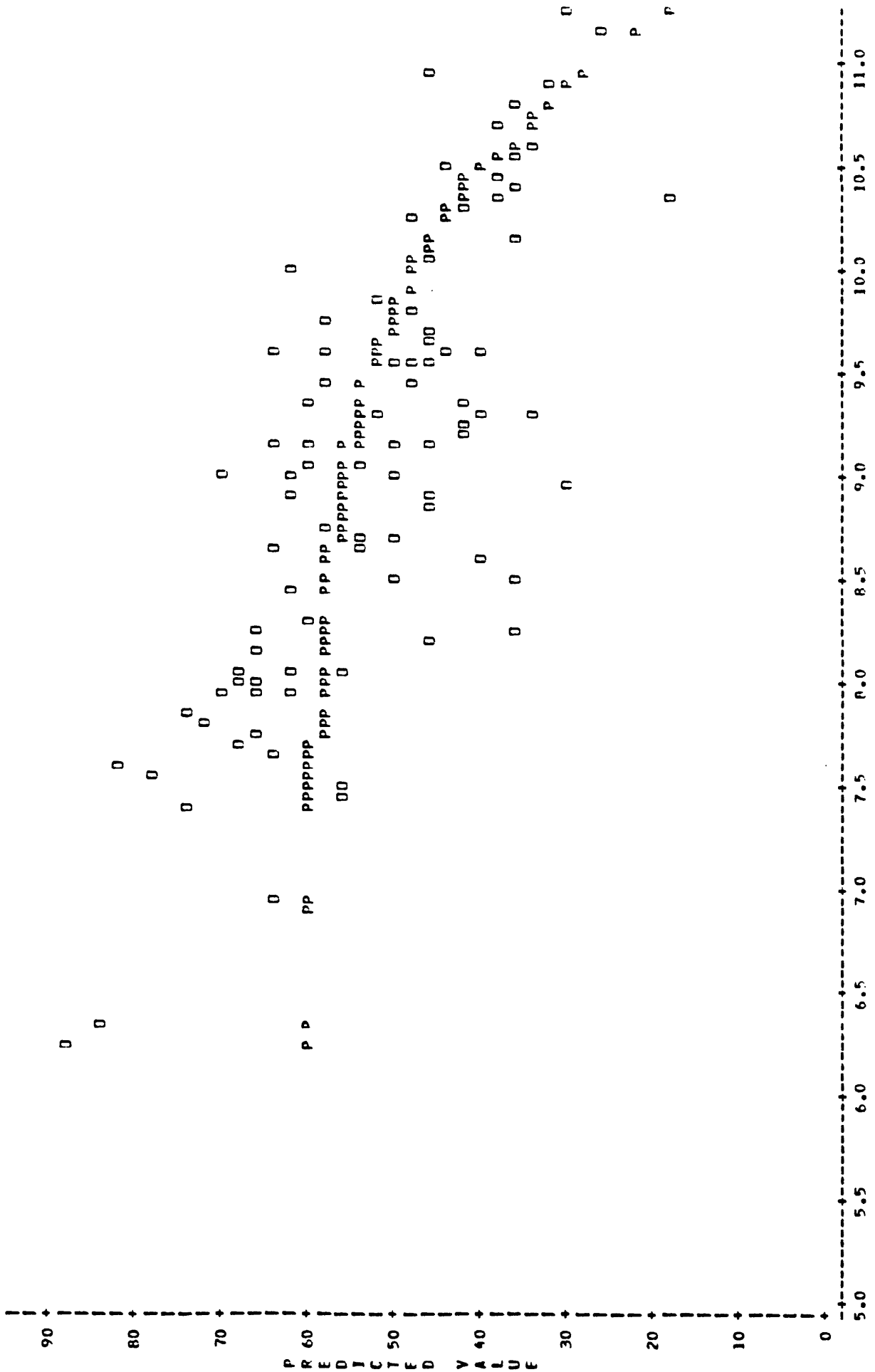
OBS	ID	ACTUAL	PREDICT VALUE	STD PREDICT ERR	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	-2-1-0 1 2	COOK'S D
31	2370.00	65.000	58.972	1.317	6.028	10.112	0.596		0.003
32	1860.00	77.000	59.233	1.336	17.767	10.109	1.758		0.027
33	2930.00	66.000	58.626	1.291	7.374	10.115	0.729		0.004
34	8379.98	60.000	55.533	1.110	4.467	10.136	0.441		0.001
35	11099.97	51.000	53.590	1.056	-2.990	10.142	-0.295		0.000
36	46600.00	33.000	33.843	2.386	-0.43347	9.914	-0.085		0.000
37	46599.91	•	33.843	2.386	•	•	•		•
38	12699.97	53.000	53.082	1.037	-0.081596	10.144	-0.008		0.000
39	3759.99	65.000	58.155	1.259	6.845	10.119	0.676		0.004
40	4899.99	50.000	57.508	1.216	-7.508	10.124	-0.742		0.004
41	3709.99	46.000	58.183	1.261	-12.183	10.119	-1.204		0.011
42	4099.99	60.000	57.962	1.246	2.038	10.121	0.201		0.000
43	1040.00	63.000	59.699	1.372	3.301	10.104	0.327		0.001
44	2790.00	70.000	58.706	1.297	11.294	10.114	1.117		0.010
45	190.00	82.000	59.165	1.331	22.835	10.110	2.259		0.044
46	514.00	88.000	59.997	1.395	28.003	10.101	2.772		0.073
47	574.00	83.000	59.963	1.393	23.037	10.101	2.281		0.049
48	25000.00	35.000	46.101	1.247	-11.101	10.170	-1.097		0.009
49	31200.00	38.000	42.583	1.521	-4.583	10.083	-0.455		0.002
50	8090.00	50.000	55.698	1.117	-5.698	10.136	-0.562		0.002
51	3100.00	56.000	58.530	1.285	-2.530	10.116	-0.250		0.001
52	3099.99	•	58.530	1.285	•	•	•		•
53	3440.00	65.000	58.337	1.271	6.663	10.117	0.659		0.003
54	1680.00	56.000	59.335	1.344	-3.335	10.108	-0.330		0.001
55	1810.00	56.000	59.262	1.338	-3.262	10.109	-0.323		0.001
56	9860.00	42.000	54.693	1.077	-12.693	10.140	-1.252		0.009
57	2840.00	66.000	58.677	1.295	7.323	10.114	0.724		0.004
58	3380.00	37.000	41.107	1.653	-4.107	10.062	-0.408		0.002
59	10700.00	54.000	54.217	1.062	-216579	10.142	-0.021		0.000
60	5620.00	64.000	57.099	1.191	6.901	10.127	0.681		0.003
61	9220.00	63.000	55.056	1.090	7.944	10.139	0.783		0.004
62	19000.00	51.000	49.506	1.073	1.494	10.140	0.147		0.000
63	17000.00	57.000	50.641	1.043	6.359	10.143	0.627		0.002
64	7320.00	63.000	56.135	1.138	6.865	10.133	0.677		0.003
65	8280.00	69.000	55.590	1.112	13.410	10.136	1.323		0.011
66	11000.00	33.000	54.046	1.057	-21.046	10.142	-2.075		0.023
67	3770.00	35.000	58.149	1.258	-23.149	10.119	-2.288		0.040
68	3180.00	62.000	58.484	1.201	3.516	10.116	0.348		0.001
69	31600.00	•	42.356	1.541	•	•	•		•
70	32200.00	36.000	42.015	1.571	-6.015	10.075	-0.597		0.004
71	8600.00	54.000	55.408	1.104	-1.408	10.137	-0.139		0.000
72	76000.00	30.000	17.159	4.253	12.841	9.268	1.386		0.202
73	30400.00	41.000	43.037	1.482	-2.037	10.089	-0.202		0.000
74	13700.00	50.000	52.514	1.032	-2.514	10.145	-0.248		0.000
75	9350.00	50.000	54.983	1.087	-4.983	10.139	-0.491		0.001
76	6820.00	45.000	51.418	1.153	-11.418	10.132	-1.127		0.008
77	6000.00	49.000	56.884	1.179	-7.884	10.129	-0.778		0.004
78	68400.00	26.000	21.472	3.760	4.528	9.478	0.478		0.018
79	41100.00	34.000	36.965	2.060	-2.965	9.987	-0.297		0.002
80	7410.00	45.000	56.084	1.136	-11.084	10.134	-1.094		0.008
81	10200.00	42.000	54.500	1.070	-12.500	10.141	-1.233		0.008
82	54900.00	31.000	29.133	2.899	1.867	9.776	0.191		0.002
83	5440.00	40.000	57.202	1.197	-17.202	10.174	-1.699		0.020

OBS	ID	ACTUAL	PREDICT VALUE	STD ERR PREDICT	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	-2-1-0 1 2	COOK'S D
84	18100.00	48.000	50.017	1.058	-2.017	10.142	-0.199		0.000
85	49800.00	35.000	32.027	2.582	2.973	9.865	0.301		0.003
86	16500.00	45.000	50.925	1.038	-5.925	10.144	-0.584	*	0.002
87	14800.00	39.000	51.890	1.030	-12.890	10.145	-1.271	**	0.008
88	9500.00	45.000	54.898	1.084	-9.898	10.139	-0.976	*	0.005
89	6000.00	.	56.884	1.179
90	7580.00	29.000	55.987	1.131	-26.987	10.134	-2.663	***	0.044
91	3200.00	56.000	58.473	1.281	-2.473	10.116	-0.244		0.000
92	2500.00	71.000	58.870	1.309	12.130	10.113	1.199	**	0.012
93	2050.00	64.000	59.125	1.328	4.875	10.110	0.482		0.002
94	3170.00	67.000	58.490	1.282	8.510	10.116	0.841	*	0.006
95	1650.00	74.000	59.352	1.345	14.648	10.108	1.449	**	0.019
96	14600.00	44.000	52.003	1.030	-8.003	10.145	-0.789	*	0.003
97	5780.00	54.000	57.009	1.186	-3.009	10.128	-0.297		0.001
98	12800.00	48.000	53.025	1.037	-5.025	10.144	-0.495		0.001
99	24100.00	45.000	46.612	1.214	-1.612	10.124	-0.159		0.000
100	11200.00	40.000	53.933	1.054	-13.933	10.142	-1.374	**	0.010
101	8160.00	62.000	55.658	1.115	6.342	10.136	0.626	*	0.002
102	4880.00	36.000	57.519	1.217	-21.519	10.124	-2.126	***	0.033
103	2860.00	61.000	58.666	1.294	2.334	10.114	0.231		0.000

SUM OF RESIDUALS 1.01608E-12
SUM OF SQUARED RESIDUALS 9981.852

OUTPJT FROM LINEAR: P00945= P00061
STATION IDENTIFICATION NUMBER=03374100

PLOT OF P*L00061 SYMBOL USED IS P
PLOT OF P00945*L00061 SYMBOL USED IS O

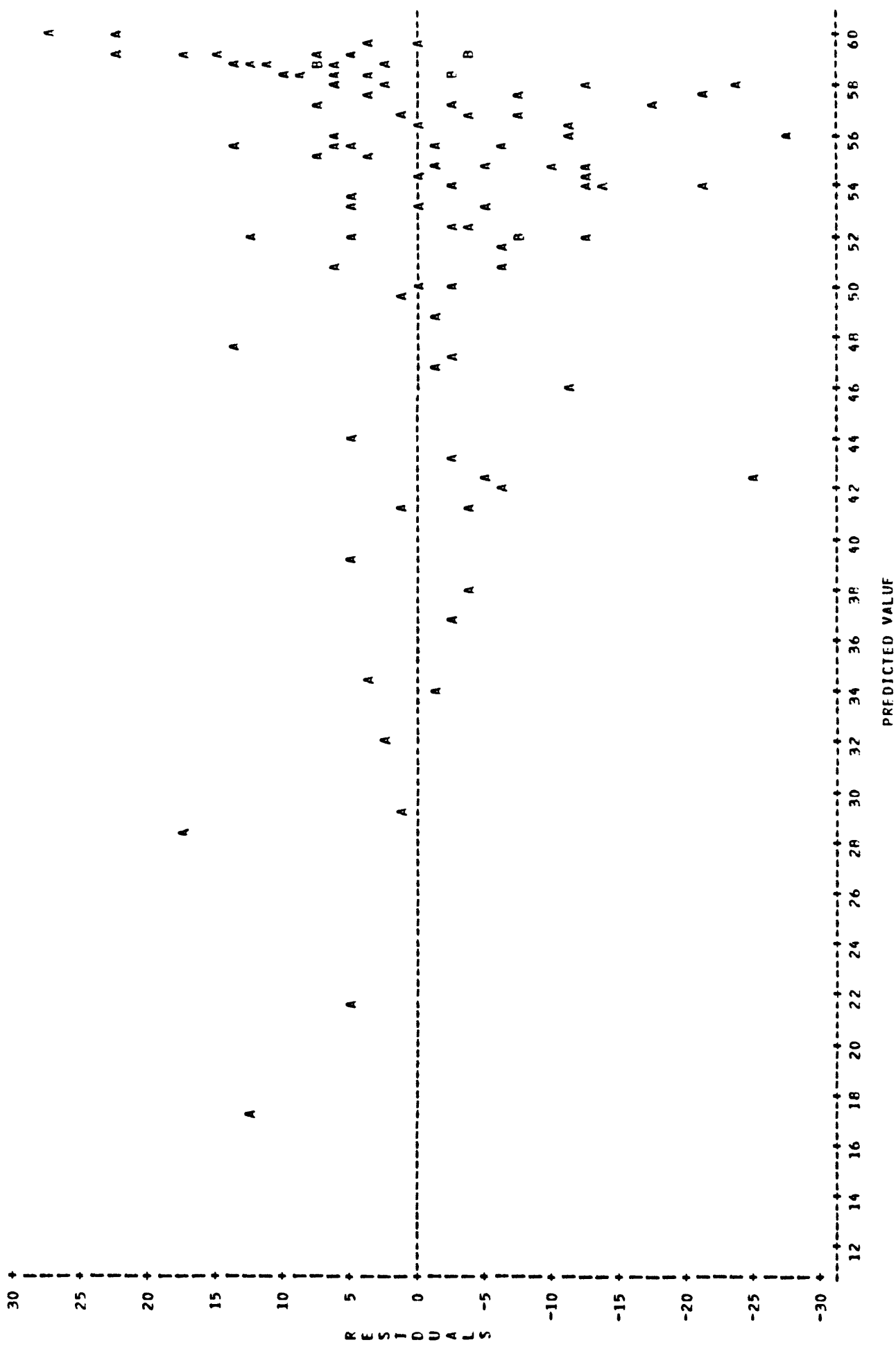


L00061

NOTE: 5 OBS HAD MISSING VALUES 47 OBS HIDDEN

OUTPUT FROM LINEAR: P00746= P0006 J
STATION IDENTIFICATION NUMBER=0337410C

PLOT OF R+P LEGEND: A = 1 OBS, B = 2 OBS, ETC.



NOTE: 5 OBS HAD MISSING VALUES

UNIVARIATE

VARIABLE=R RESIDUALS

MOMENTS

N 9R
 MEAN 1.037E-14 SUM MGTS 9R
 STD DEV 10.1442 VARIANCE 1.016E-12
 SKEWNESS -0.130921 KURTOSIS 0.660651
 USS 9981.85 CSS 9981.85
 CV 9.784E+16 STD MEAN 1.02472
 T:MEAN=0 1.012E-14 PROR>|T| 1
 SGN RANK 69.5 PROR>|S| 0.80683R
 NUM ^= 0 9P
 D:NORMAL 0.0652829 PROR>> >0.15

QUANTILES(DEF=4)

100% MAX 28.0029 99% 28.0029
 75% Q3 6.34615 95% 17.4509
 50% MED -0.149088 90% 12.89R
 25% Q1 -5.19306 10% -12.7116
 0% MIN -26.9872 5% -21.07
 1% -26.9872

EXTREMES

LOWEST -26.9872 HIGHEST 17.4343
 -25.4694 17.7667
 -23.1493 22.8348
 -21.5194 23.0369
 -21.0463 28.0029

RANGE 54.99
 Q3-Q1 11.5392
 MODE -26.9872

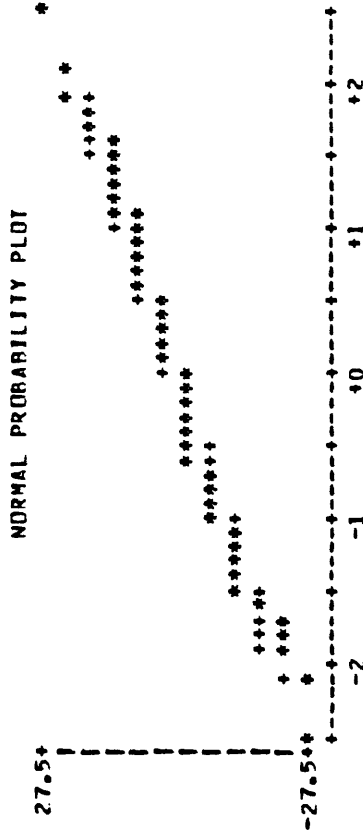
MISSING VALUE .
 COUNT 5
 % COUNT/NOBS 4.85

BOXPLOT

```

STEM LEAF          #
 2 8                1
 2 33               2
 1 578              3
 1 1223334         7
 0 555555666777778899 20
 0 1112223333444 13
 -0 44433333332222211110000 25
 -0 88877666555 11
 -1 4333321110 10
  -1 7             1
  -2 321          3
  -2 75          2
  -----+-----+
  MULTIPLY STEM=LEAF BY 10**+01
    
```

NORMAL PROBABILITY PLOT



MODEL: LOGLIN
DEP VARIABLE: P00945 SULFATE DISSOLVED (MG/L AS SO4)

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB>F
MODEL	1	9971.154	9971.154	128.648	0.0001
ERROR	96	7440.682	77.507101		
C TOTAL	97	17411.835			

ROOT MSE 6.803812
DEP MEAN 52.040815
C.V. 16.91713

PARAMETER ESTIMATE STANDARD ERROR T FOR HO: VARIABLE LABEL

INTERCFP 1 136.455 7.495376 18.205 0.0001 INTERCEPT
L0006J 1 -9.317384 0.821471 -11.342 0.0001

OBS	ID	ACTUAL	PREDICT VALUE	STD ERR PREDICT	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	COOK'S D
1	12799.97	58.000	48.339	0.947318	9.661	9.753	1.104	0.007
2	55899.89	46.000	34.604	1.776	11.396	8.623	1.322	0.037
3	28399.94	49.000	40.913	1.324	8.086	8.704	0.929	0.010
4	9299.98	59.000	51.315	0.891618	7.685	8.759	0.877	0.004
5	15299.97	45.000	46.677	1.007	-1.677	8.746	-0.192	0.000
6	31399.93	17.000	39.978	1.386	-22.978	8.694	-2.643	0.089
7	5989.99	53.000	55.414	0.937731	-2.414	8.754	-0.276	0.000
8	3029.99	68.000	61.764	1.235	6.236	8.717	0.715	0.005
9	2750.00	67.000	64.537	1.416	2.463	9.699	0.283	0.001
10	2600.00	73.000	63.190	1.326	9.810	8.703	1.127	0.015
11	6360.00	58.000	54.856	0.923297	3.144	8.755	0.359	0.001
12	14799.96	57.000	46.986	0.994729	10.014	8.747	1.145	0.028
13	22500.00	61.000	43.083	1.189	17.917	8.723	2.054	0.039
14	33599.93	42.000	39.347	1.429	2.653	8.687	0.305	0.001
15	37000.00	44.000	38.449	1.492	5.551	8.676	0.640	0.006
16	14100.00	48.000	47.438	0.977546	0.562374	8.679	0.064	0.000
17	11700.00	59.000	49.176	0.924488	9.824	8.755	1.122	0.007
18	4740.00	61.000	57.595	1.015	3.405	8.745	0.389	0.001
19	7979.98	41.000	52.741	0.891462	-8.500	8.756	-0.971	0.005
20	11300.00	45.000	47.306	0.982405	-2.306	8.749	-0.264	0.000
21	14299.96	45.000	42.798	1.206	2.202	8.721	0.253	0.001
22	23199.95	45.000	44.042	1.135	3.958	8.730	0.453	0.002
23	20299.95	48.000	37.958	1.527	-2.958	8.670	-0.341	0.002
24	38999.92	35.000	45.111	1.079	4.889	8.737	0.560	0.002
25	18099.96	50.000	36.563	1.629	1.437	8.652	0.166	0.000
26	45299.91	38.000	46.923	0.997220	17.077	8.747	1.952	0.025
27	14899.97	64.000	50.913	0.894859	2.087	8.758	0.238	0.000
28	9709.98	53.000	54.581	0.917093	1.419	8.756	0.162	0.000
29	6549.99	56.000	72.281	1.994	-17.281	8.575	-1.432	0.055
30	980.00	60.000						

OBS	ID	ACTUAL	PREDICT VALUE	STD ERR PREDICT	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	-2-1-0 1 2	COOK'S D
31	2320.00	65.000	64.252	1.396	0.748136	8.692	0.086		0.000
32	1860.00	77.000	66.311	1.541	10.689	8.668	1.233	**	0.024
33	2930.00	66.000	62.077	1.255	3.923	8.714	0.450		0.002
34	8379.9R	60.000	52.286	0.849581	7.714	8.759	0.881	*	0.004
35	11099.97	51.000	49.667	0.913621	1.333	8.756	0.152		0.000
36	46600.00	33.000	36.299	1.648	-3.299	8.648	-0.382		0.003
37	46599.91	53.000	48.412	0.94511R	4.508	8.753	0.524	*	0.002
38	12699.97	65.000	59.753	1.119	5.247	8.732	0.601	*	0.003
39	3759.99	50.000	57.286	1.002	-7.286	8.747	-0.833	*	0.005
40	4899.99	46.000	59.878	1.126	-13.878	8.731	-1.589	***	0.021
41	3709.99	60.000	58.946	1.078	1.054	8.73P	0.121		0.000
42	4099.99	63.000	71.728	1.950	-8.728	8.585	-1.017	*	0.027
43	1040.00	70.000	62.533	1.283	7.467	8.710	0.857		0.008
44	2790.00	82.000	65.728	1.499	16.272	8.675	1.876	***	0.053
45	1980.00	88.000	78.294	2.480	9.706	8.447	1.149	***	0.057
46	514.00	83.000	77.265	2.395	5.735	8.472	0.677	*	0.018
47	574.00	35.000	42.102	1.249	-7.102	9.715	-0.815	*	0.007
48	25000.00	38.000	40.037	1.382	-2.037	8.695	-0.234		0.001
49	31200.00	50.000	52.614	0.890753	-2.614	8.759	-0.298		0.000
50	8090.00	56.000	61.551	1.222	-5.551	8.719	-0.637	*	0.004
51	3100.00	65.000	61.551	1.222	4.418	8.726	0.506	*	0.002
52	3099.99	56.000	67.259	1.610	-11.259	8.655	-1.301	**	0.029
53	3440.00	56.000	66.565	1.559	-10.565	8.665	-1.219	**	0.024
54	1680.00	42.000	50.770	0.896346	-8.770	8.758	-1.001	**	0.005
55	1810.00	66.000	62.368	1.273	3.632	8.711	0.417		0.002
56	9860.00	37.000	39.292	1.433	-2.292	9.686	-0.264		0.001
57	2840.00	54.000	50.009	0.907189	3.991	9.757	0.456		0.001
58	33800.00	64.000	56.008	0.955634	7.992	9.752	0.913	*	0.005
59	10700.00	63.000	51.396	0.891137	11.604	8.735	1.325	**	0.009
60	5620.00	51.000	44.659	1.102	6.341	8.741	0.726	**	0.004
61	9220.00	57.000	45.695	1.051	11.305	8.741	1.293	**	0.012
62	19000.00	63.000	53.546	0.899163	9.454	8.758	1.080	**	0.036
63	17000.00	69.000	52.398	0.889875	16.602	8.759	1.896	***	0.019
64	7320.00	33.000	49.751	0.911947	-16.751	8.756	-1.913	***	0.020
65	8280.00	35.000	59.728	1.11R	-24.728	8.733	-2.832	****	0.066
66	10000.00	62.000	61.314	1.208	0.686054	8.721	0.079		0.000
67	3770.00	36.000	39.919	1.390	-3.743	8.691	-0.431		0.002
68	3180.00	54.000	52.044	0.889319	1.956	8.759	0.223		0.000
69	31600.00	30.000	31.742	1.998	-1.742	8.574	-0.203		0.001
70	32200.00	41.000	40.279	1.366	0.720620	8.697	0.083		0.000
71	8600.00	50.000	47.706	0.97970	2.294	8.750	0.262		0.000
72	76000.00	50.000	51.265	0.891945	-1.265	9.759	-0.144		0.000
73	30400.00	45.000	54.205	0.909558	-9.205	8.757	-1.051	**	0.006
74	13700.00	49.000	55.399	0.937297	-6.399	8.754	-0.731	*	0.033
75	9350.00	26.000	32.724	1.921	-6.724	8.592	-0.783	*	0.015
76	6820.00	34.000	37.470	1.562	-3.470	8.664	-0.400		0.033
77	4100.00	45.000	53.432	0.897737	-8.432	8.758	-0.903	*	0.005
78	7410.00	42.000	50.454	0.900250	-8.454	8.758	-0.965	*	0.005
79	10200.00	31.000	34.772	1.763	-3.772	8.625	-0.437		0.004
80	54900.00	40.000	56.311	0.965742	-16.311	8.751	-1.864	***	0.021

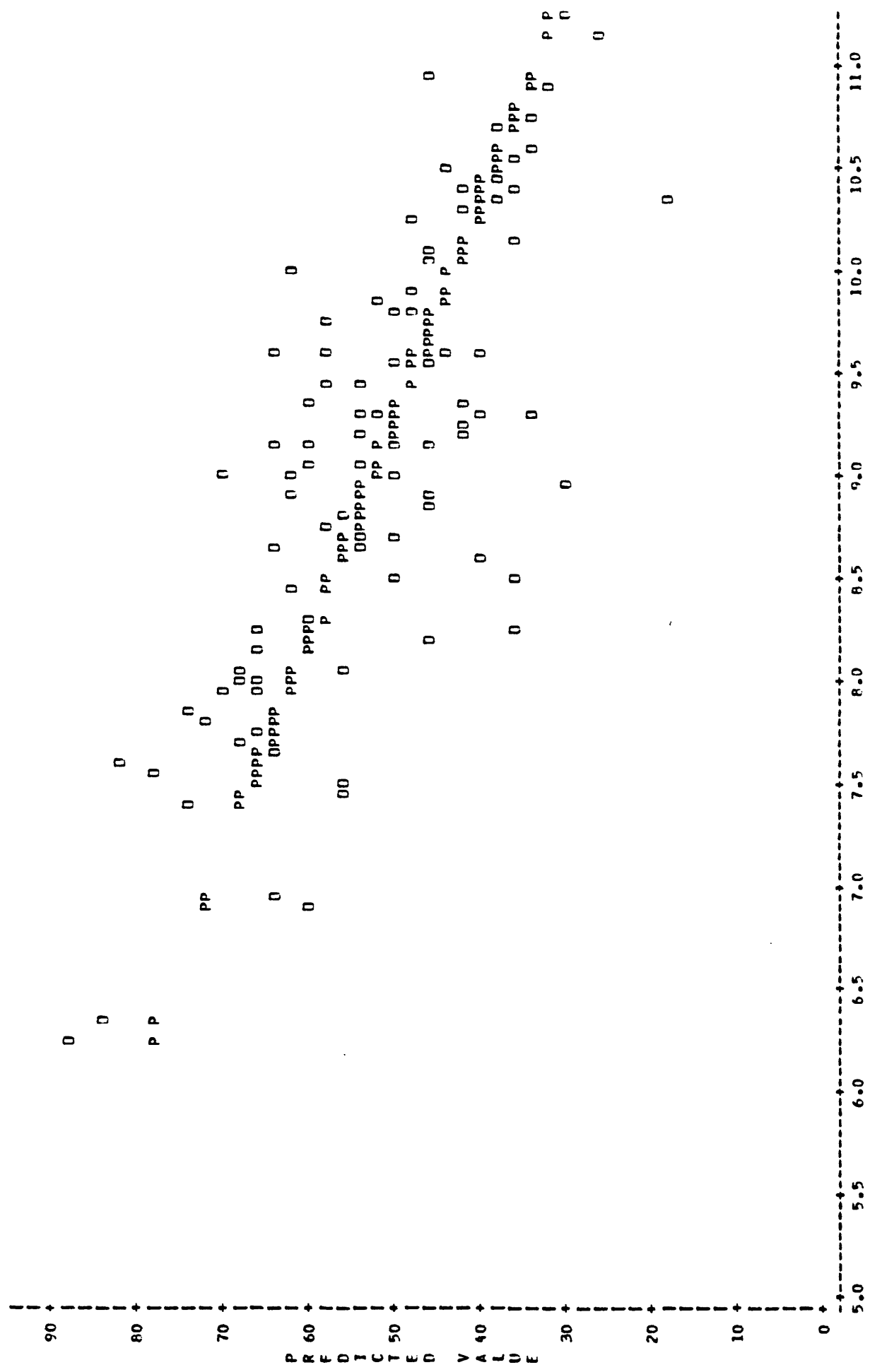
OUTPUT FROM LOGLIN: P00945= L0006 J
STATION IDENTIFICATION NUMBER=03374100

OBS	ID	ACTUAL	PREDICT VALUE	STD ERR PREDICT	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	-2-1-0 1 2	CORR'S D
84	18100.00	48.000	45.111	1.079	2.889	8.737	0.331		0.001
85	49800.00	35.000	35.681	1.695	-680576	8.639	-0.079		0.000
86	16500.00	45.000	45.973	1.038	-.973067	9.742	-0.111		0.000
87	14800.00	39.000	46.986	0.994730	-7.986	8.747	-0.913	*	0.005
88	9500.00	45.000	51.117	0.893042	-6.117	8.758	-0.698	*	0.003
89	6000.00	*	55.399	0.937297	*	*	*		*
90	7580.00	29.000	53.221	0.995381	-24.221	8.758	-2.765	***	0.040
91	3200.00	56.000	61.256	1.205	-5.256	8.721	-0.603	*	0.003
92	2500.00	71.000	63.556	1.350	7.444	8.700	0.856	*	0.009
93	2050.00	64.000	65.405	1.476	-1.405	8.679	-0.162		0.000
94	3170.00	67.000	61.343	1.210	5.657	8.720	0.649	*	0.004
95	1650.00	74.000	67.427	1.622	6.573	8.653	0.760	*	0.010
96	14600.00	44.000	47.113	0.979773	-3.113	8.748	-0.356		0.031
97	5780.00	54.000	55.747	0.947437	-1.747	8.753	-0.200		0.000
98	12800.00	48.000	48.339	0.947319	-.338893	9.753	-0.039		0.000
99	24100.00	45.000	42.443	1.228	2.557	8.718	0.293		0.001
100	11200.00	40.000	49.583	0.915338	-9.583	8.756	-1.094	**	0.007
101	8160.00	62.000	52.534	0.890380	9.466	8.759	1.081	**	0.006
102	4880.00	36.000	57.324	1.004	-21.324	8.746	-2.438	***	0.039
103	2860.00	61.000	62.302	1.269	-1.302	8.712	-0.149		0.000

SUM OF RESIDUALS 2.07905E-11
SUM OF SQUARED RESIDUALS 7440.682

OUTPUT FROM LOGLIN: P00945= L00061
STATION IDENTIFICATION NUMBER=03374100

PLOT OF P*L00061 SYMBOL USED IS P
PLOT OF P00945*L00061 SYMBOL USED IS 0

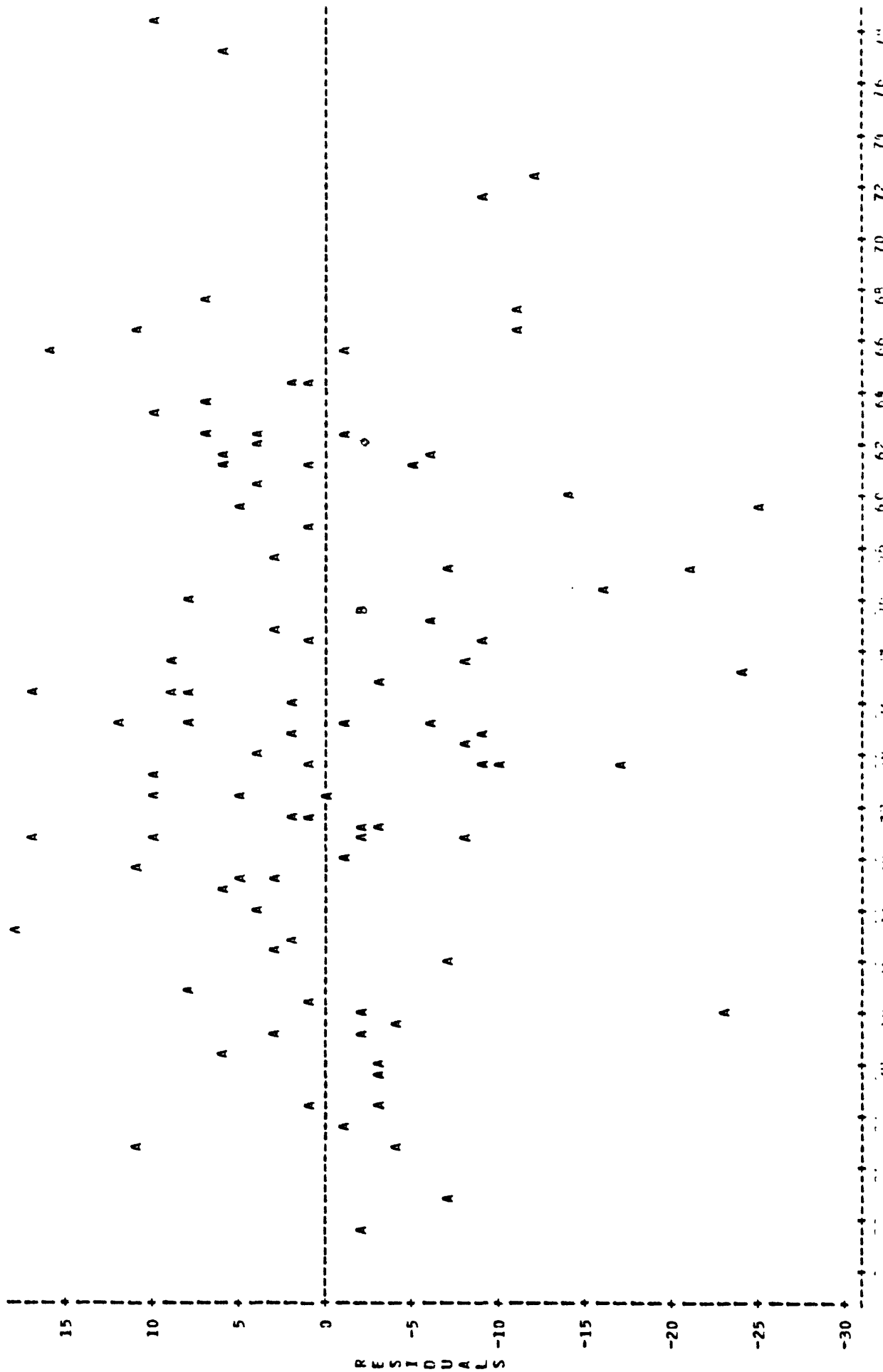


L00061

NOTE: 5 OBS HAD MISSING VALUES 44 OBS HIDDEN

OUTPUT FROM LOGLIN: P00945= L00061
STATION IDENTIFICATION NUMBER=033374100

PLOT OF R*P LEGEND: A = 1 OBS, B = 2 OBS, ETC.



UNIVARIATE

VARIABLE=R RESIDUALS

MOMENTS

N 98
 MEAN 2.121E-13
 STD DEV 8.75831
 SKEWNESS -0.59734E
 USS 7440.68
 CV 4.128E+15
 T:MEAN=0
 SGN RANK 2.399E-13
 NUM ^= 0 167.5
 D:NORMAL 0.0782438

QUANTILES (DEF=4)

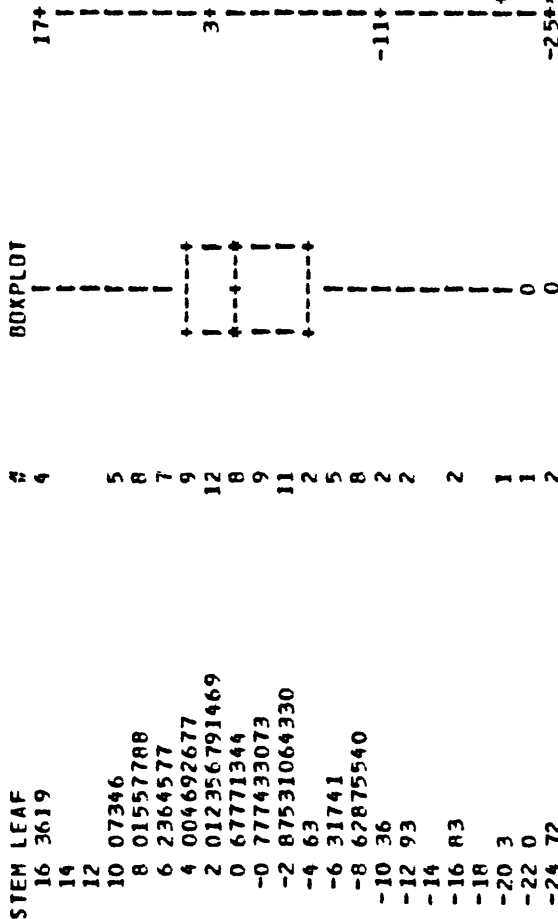
100% MAX 17.916R
 75% Q3 5.85993
 50% MED 0.9008R6
 25% Q1 -5.3294R
 0% MIN -24.72R2
 RANGE 42.645
 Q3-Q1 11.1894
 MODE -24.72R2

EXTREMES

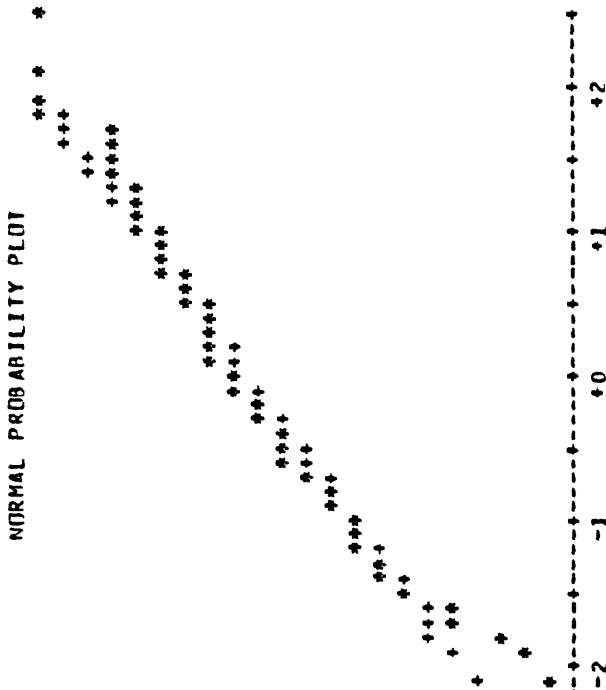
LOWEST 17.916R
 -24.72R2
 -24.2206
 -22.9778
 -21.3236
 -16.7509
 HIGHEST 11.6044
 16.2716
 16.6024
 17.0765
 17.9168

MISSING VALUE .
 COUNT 5
 % COUNT/NOBS 4.85

BOXPLOT



NORMAL PROBABILITY PLOT



OUTPUT FROM LOGLOG: L00945= L00061
STATION IDENTIFICATION NUMBER=03374100

MODEL: LOGLOG
DEP VARIABLE: L00945

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROR>F
MODEL	1	3.883952		104.235	0.0001
ERROR	96	3.577098			
C TOTAL	97	7.461050			
ROOT MSE		0.193032	R-SQUAFE	0.5205	
DEP MEAN		3.916335	ADJ R-SQ	0.5155	
C.V.		4.928899			

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T
INTERCEP	1	5.582358	0.164343	33.968	0.0001
L00061	1	-0.183890	0.018012	-10.219	0.0001

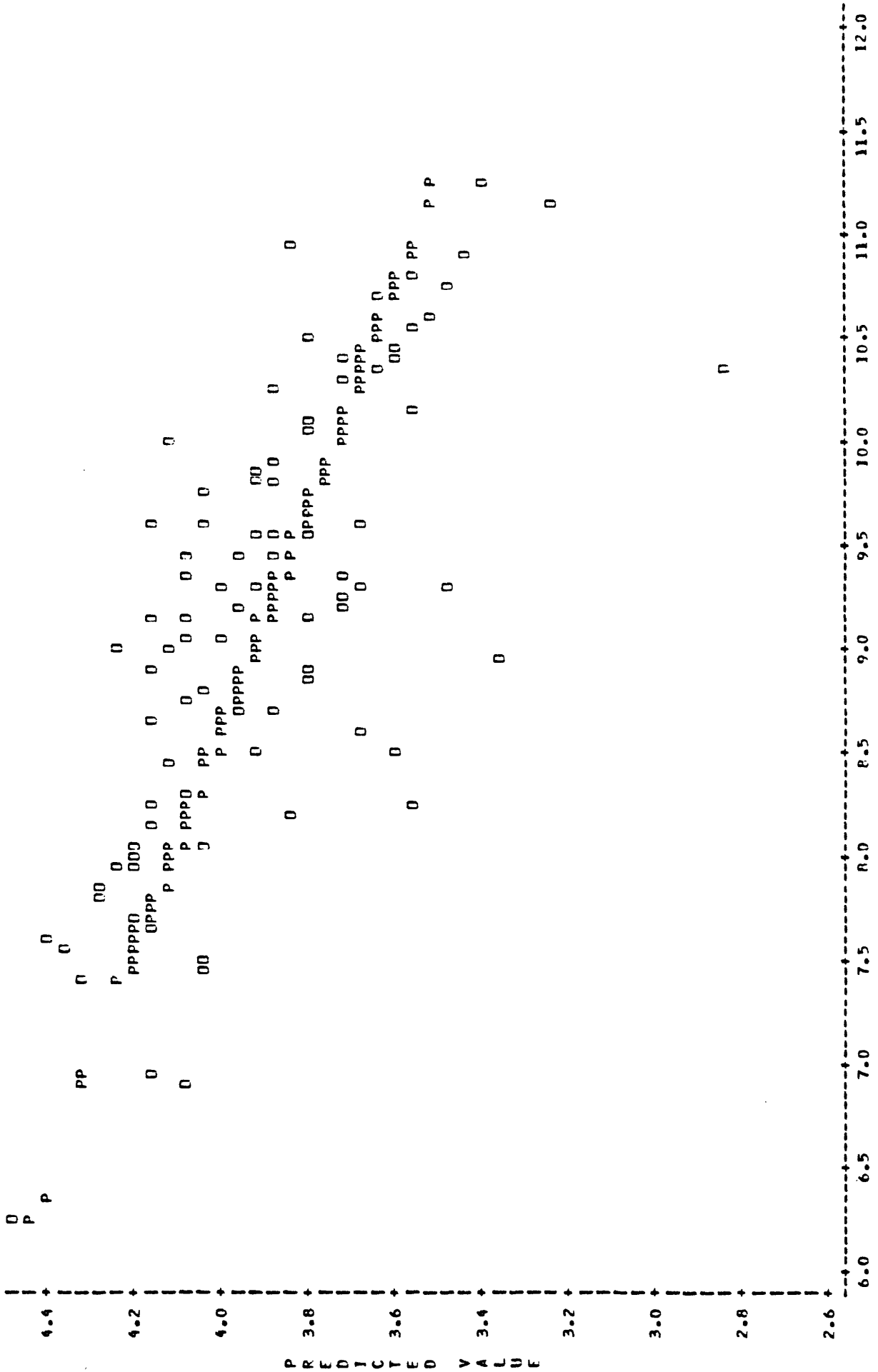
OBS	ID	ACTUAL	PREDICT VALUE	STD ERR PREDICT	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	-2-1-0 1 2	COOK'S D
1	12799.97	4.060	3.843	0.020771	0.217170	0.191911	1.132	**	0.008
2	55899.89	3.829	3.572	0.031941	0.256444	0.189064	1.356	**	0.039
3	28399.94	3.892	3.697	0.029033	0.195096	0.190836	1.022	**	0.012
4	9299.98	4.078	3.902	0.019550	0.175524	0.192040	0.914	*	0.004
5	15299.97	3.807	3.810	0.022095	-0.003804	0.191765	-0.020		0.000
6	31399.93	2.833	3.678	0.030397	-0.845044	0.190624	-4.433	*****	0.250
7	5989.99	3.970	3.983	0.020561	-0.12619	0.191934	-0.066		0.000
8	3029.99	4.220	4.108	0.027084	0.111270	0.191123	0.582	*	0.003
9	2250.00	4.205	4.136	0.029065	0.041724	0.190519	0.219	*	0.001
10	2600.00	4.060	3.972	0.020244	0.088554	0.191968	0.461		0.001
11	6360.00	4.043	3.817	0.021810	0.226475	0.191796	1.181	**	0.009
12	14799.96	4.111	3.740	0.026078	0.371328	0.191263	1.941	***	0.035
13	22500.00	4.111	3.666	0.031343	0.071865	0.190471	0.377		0.002
14	33599.93	3.738	3.648	0.032720	0.136111	0.190239	0.715	*	0.008
15	37000.00	3.784	3.825	0.021434	0.045716	0.191839	0.238		0.000
16	14100.00	3.871	3.860	0.020270	0.217741	0.191965	1.134	**	0.007
17	11700.00	4.078	4.026	0.022260	0.084923	0.191744	0.443		0.001
18	4740.00	4.111	3.930	0.019546					0.000
19	7979.98		3.866	0.020108	-1.52621	0.191982	-0.795	*	0.003
20	11300.00	3.714	3.823	0.021540	-0.162233	0.191827	-0.085		0.000
21	14299.96	3.807	3.734	0.024448	0.072750	0.191212	0.380		0.001
22	23199.95	3.871	3.758	0.024886	0.112733	0.191421	0.589	*	0.003
23	20299.95	3.871	3.638	0.033486	-0.083050	0.190106	-0.437		0.003
24	38999.92	3.555	3.780	0.023658	0.132461	0.191577	0.691	*	0.004
25	18099.96	3.912	3.611	0.035713	0.026724	0.189700	0.141		0.000
26	45299.91	3.638	3.815	0.021865	0.343546	0.191790	1.791	***	0.021
27	14899.97	4.159	3.894	0.019621	0.076212	0.192032	0.397		0.001
28	9709.98	3.970	3.966	0.020108	0.058876	0.191932	0.307		0.001
29	6549.99	4.025	4.316	0.043717	-0.221461	0.188017	-1.178	**	0.038
30	980.00	4.094							

OUTPUT FROM LOGLOG: L00945= L00061
STATION IDENTIFICATION NUMBER=03374100

OBS	ID	ACTUAL	PREDICT VALUE	STD ERR PREDICT	RESIDUAL	STD ERR RESIDUAL	STUDENT RESIDUAL	-2-1-0 1 2	CORR'S D
31	2320.00	4.174	4.157	0.030617	0.017052	0.190589	0.089		0.000
32	1860.00	4.344	4.198	0.033782	0.145832	0.190053	0.767		0.009
33	2930.00	4.190	4.114	0.027507	0.075246	0.191052	0.394	*	0.002
34	8379.98	4.094	3.921	0.019505	0.173176	0.192044	0.902	*	0.004
35	11099.97	3.932	3.869	0.020032	0.062340	0.191990	0.325		0.001
36	46600.00	3.497	3.606	0.036141	-0.109151	0.189619	-0.576	*	0.006
37	46599.91	•	3.606	0.036141	•	•	•		•
38	12699.97	3.970	3.845	0.020723	0.125576	0.191917	0.654	*	0.002
39	3759.99	4.174	4.069	0.024545	0.105844	0.191465	0.553	*	0.003
40	4899.99	3.912	4.020	0.021978	-0.107824	0.191777	-0.562	*	0.002
41	3709.99	3.829	4.071	0.024693	-0.242364	0.191446	-1.266	**	0.013
42	4099.99	4.094	4.053	0.023631	0.041720	0.191580	0.218		0.000
43	1040.00	4.143	4.305	0.042761	-0.161744	0.188236	-0.859	*	0.019
44	2790.00	4.248	4.123	0.028135	0.125083	0.190971	0.655	*	0.005
45	1980.00	4.407	4.186	0.032868	0.220243	0.190213	1.158	**	0.020
46	514.00	4.477	4.434	0.054368	0.042861	0.185218	0.231		0.002
47	574.00	4.419	4.414	0.052516	0.004668	0.185751	0.025		0.000
48	25000.00	3.555	3.720	0.027375	-0.164823	0.191081	-0.863	*	0.008
49	31200.00	3.638	3.679	0.030309	-0.41846	0.190638	-0.220		0.001
50	8090.00	3.912	3.928	0.019531	-0.015622	0.192042	-0.081		0.000
51	3100.00	4.025	4.104	0.026800	-0.078685	0.191163	-0.412		0.002
52	3099.99	•	4.104	0.026800	•	•	•		•
53	3440.00	4.174	4.085	0.025550	0.089488	0.191334	0.468		0.002
54	1680.00	4.025	4.217	0.035294	-0.191338	0.189778	-1.008	**	0.018
55	1810.00	4.025	4.203	0.034183	-0.177632	0.189981	-0.935	*	0.014
56	9860.00	3.738	3.891	0.019653	-0.153591	0.192029	-0.800	*	0.003
57	2840.00	4.190	4.120	0.027906	0.069509	0.191004	0.364		0.001
58	33800.00	3.611	3.665	0.031427	-0.053795	0.190457	-0.282		0.001
59	10700.00	3.989	3.876	0.019891	0.112758	0.192005	0.587	*	0.002
60	5620.00	4.159	3.995	0.020953	0.164248	0.191892	0.856	*	0.004
61	9220.00	4.143	3.904	0.019539	0.239533	0.192041	1.247	**	0.008
62	19000.00	3.932	3.771	0.024163	0.161188	0.191514	0.842	*	0.006
63	17000.00	4.043	3.791	0.023037	0.251960	0.191653	1.315	**	0.012
64	7320.00	4.143	3.946	0.019715	0.197096	0.192023	1.026	**	0.006
65	8280.00	4.234	3.923	0.019511	0.310731	0.192044	1.618	**	0.014
66	11000.00	3.497	3.871	0.019995	-0.374634	0.191994	-1.951	***	0.021
67	3770.00	3.555	4.068	0.024516	-0.512707	0.191469	-2.678	****	0.059
68	3180.00	4.127	4.099	0.026487	0.027783	0.191206	0.145		0.000
69	31600.00	•	3.677	0.030485	•	•	•		•
70	32200.00	3.584	3.674	0.030746	-0.090112	0.190568	-0.473		0.003
71	8600.00	3.989	3.916	0.019499	0.072581	0.192045	0.378		0.001
72	76000.00	3.401	3.516	0.043818	-0.114514	0.187993	-0.609	*	0.010
73	30400.00	3.714	3.684	0.029953	0.029363	0.190694	0.154		0.000
74	13700.00	3.912	3.831	0.021224	0.081245	0.191852	0.423		0.001
75	9350.00	3.912	3.901	0.019557	0.010996	0.192039	0.057		0.000
76	6820.00	3.807	3.959	0.019943	-0.152385	0.191999	-0.794	*	0.003
77	6000.00	3.892	3.983	0.020551	-0.090784	0.191935	-0.473	*	0.001
78	68400.00	3.258	3.535	0.042127	-0.276990	0.188379	-1.470	**	0.054
79	41100.00	3.526	3.629	0.034259	-0.102393	0.189968	-0.539	*	0.005
80	7410.00	3.807	3.944	0.019684	-0.137128	0.192026	-0.714	*	0.003
81	10200.00	3.738	3.885	0.019739	-0.147357	0.192020	-0.767	*	0.003
82	54900.00	3.434	3.576	0.038660	-0.141529	0.189121	-0.748	*	0.012
83	5440.00	3.689	4.001	0.0271175	-0.311747	0.191847	-1.625	**	0.016

OUTPUT FROM LOGLOG: L00945= L00061
STATION IDENTIFICATION NUMBER=03374100

PLOT OF P*L00061 SYMBOL USED IS P
PLOT OF L00945*L00061 SYMBOL USED IS O



L00061

NOTE: 5 OBS HAD MISSING VALUES 45 OBS HIDDEN

UNIVARIATE

VARIABLE=R RESIDUALS

MOMENTS

N 98
 MEAN 1.382E-14
 STD DEV 0.192035
 SKEWNESS -1.35597
 USS 3.5771
 CV 1.390E+15
 T:MEAN=0
 SGN RANK 7.124E-13
 NUM T= 0
 D:NORMAL 0.0989695

SUM WGTs 9R
 SUM 1.354E-12
 VARIANCE 0.0368773
 KUPTOSIS 3.74112
 CSS 3.5771
 STD MEAN 0.0193984
 PROB>|T| 1
 PROB>|S| 0.368084
 PROB>D 0.01R

QUANTILES (DEF=4)

100% MAX 0.371328
 75% Q3 0.11434
 50% MED 0.0286457
 25% Q1 -0.103751
 OR MIN -0.845044

RANGF 1.21637
 Q3-Q1 0.21809
 MODE -0.845044

99% 0.371328
 95% 0.252185
 90% 0.217227
 10% -0.19435
 5% -0.377756
 1% -0.845044

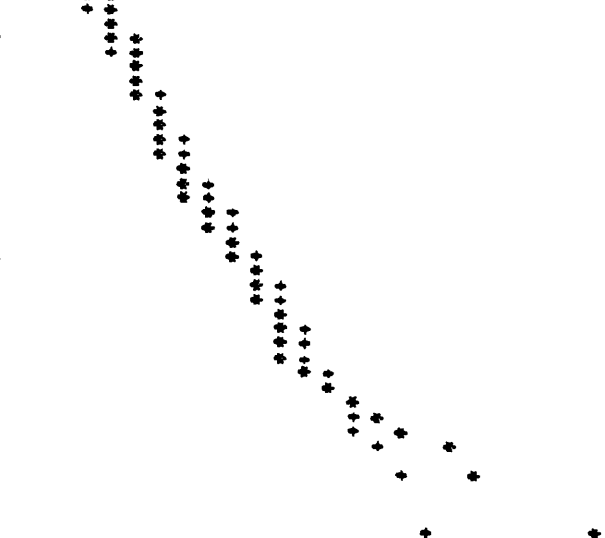
EXTREMES
 LDWFS -0.845044
 HIGHEST 0.25196
 0.256444
 0.310731
 0.343545
 0.371328

MISSING VALUE
 COUNT 5
 % COUNT/NBS 4.85

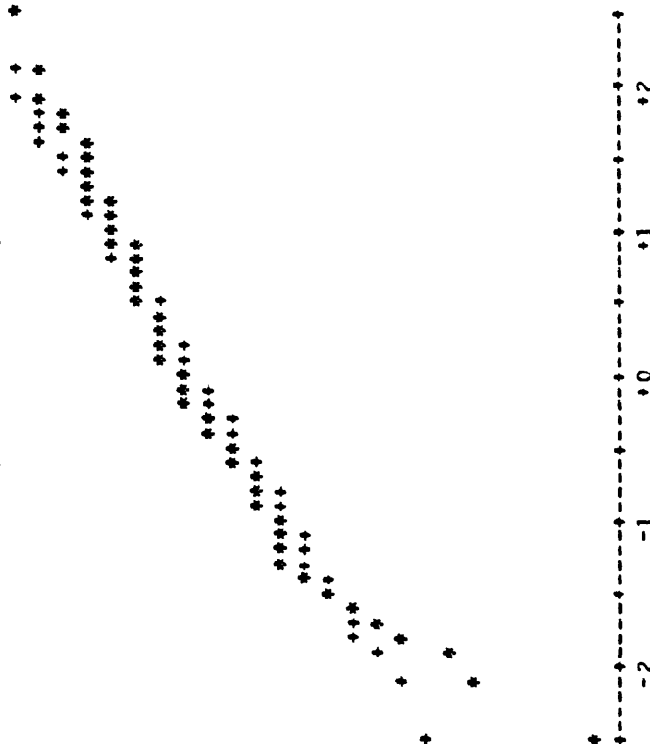
STEM LEAF

3	7	1	0
3	14	2	0
2	56	2	0
2	00022234	8	0
1	556678	6	0
1	0111123334	10	0
0	5667777888888999	16	0
0	1123333444	10	0
-0	4322211000	11	0
-0	9998875	7	0
-1	441110	6	0
-1	9886655555	10	0
-2	42	2	0
-2	R	1	0
-3	1	1	0
-3	7	1	0
-4	4	1	0
-4		1	0
-5	1	1	0
-5	7	1	0
-6			
-6			
-7			
-7			
-8	5	1	0

ROXPLOT



NORMAL PROBABILITY PLOT



MULTIPLY STEM.LEAF BY 10**+01

tools which aid the user in selecting the most appropriate regression model. The program uses the macro REGDATA to establish the data. The macro REGMAC uses the REG procedure (SAS Institute, Inc., 1982b, p. 39) rather than the SYSREG procedure to generate regression results. In addition to the information provided by SYSREG, the REG procedure in this application prints (1) the observed and predicted value for each observation; (2) the residual (FAC), the standard error of the residual and the studentized residual (the residual divided by its standard error); and (3) Cook's D, a measure of the change to the parameter estimates that would result from deleting each observation. This program also plots for each model the observed and predicted values of the selected dependent variables against the log of streamflow, and the FAC (residuals) against the predicted value. In these plots, the log of the flow is used on the x-axis to provide better resolution in the low discharge range. Finally, the macro does a residuals analysis with the Univariate procedure (SAS Institute, Inc., 1979, p. 427). This procedure does univariate statistics on the residuals, tests for normality by the Shapiro-Wilk W statistic (for sample sizes ≤ 50) or a modified version of the Kolmogorov-Smirnov D Statistic (for sample sizes > 50) and provides a stemleaf, box plot, and probability plot of the residuals. For a discussion of regression analysis and the use of these diagnostic tools, the reader is referred to Chatterjee and Price (1977), Belsley and others (1980), Daniel and Wood (1980), and Draper and Smith (1981).

The procedure is invoked three times in the example, shown in figure 4, once for the linear model of equation 2 (lines 65-68), once for the log-linear model of equation 3 (lines 69-72), and for the log-log model of equation 7 (lines 73-76). In each instance three variables are set - the model label (MODLABEL) in lines 65, 69, and 73, the independent variable (Y) in lines 66, 70, and 74, and the explanatory variables (FLOW) in lines 67, 71, and 75 -

then the macro REGMAC is called (lines 68, 72, and 76) to do the computation.

In the example shown in figure 4, the log-linear (equation 3) was selected to flow adjust the sulfate concentrations. The linear model (equation 2) is clearly deficient. Both plots demonstrate this quite clearly. However, the log-log model (equation 7) is not manifestly better or worse than the log-linear model. The pattern seen on the residuals plots for these two models are very similar, both showing some modest lack of fit at the low end and a slight amount of heteroscedasticity. The choice between these two models, in this case, is largely a matter of preference. The log-linear model has the advantage that the residuals are in units of mg/L.

PLOTTING WATER-QUALITY DATA AS A TIME SERIES

It is now possible to examine the flow-adjusted concentrations for trend. A good exploratory method for detecting trends is to plot the data as a time series. Figure 5 shows an example SAS program that generates time series plots for streamflow, sulfate concentration, and the flow-adjusted sulfate concentration for the data used in figure 3. Note that PROC REG (lines 12-14) is used to calculate the FAC and generate the data set used by PROC PLOT (lines 16-19) (SAS Institute, Inc., 1979, p. 343). Note also that the variable DECTIME (lines 17-19) is used as the time variable in these plots.

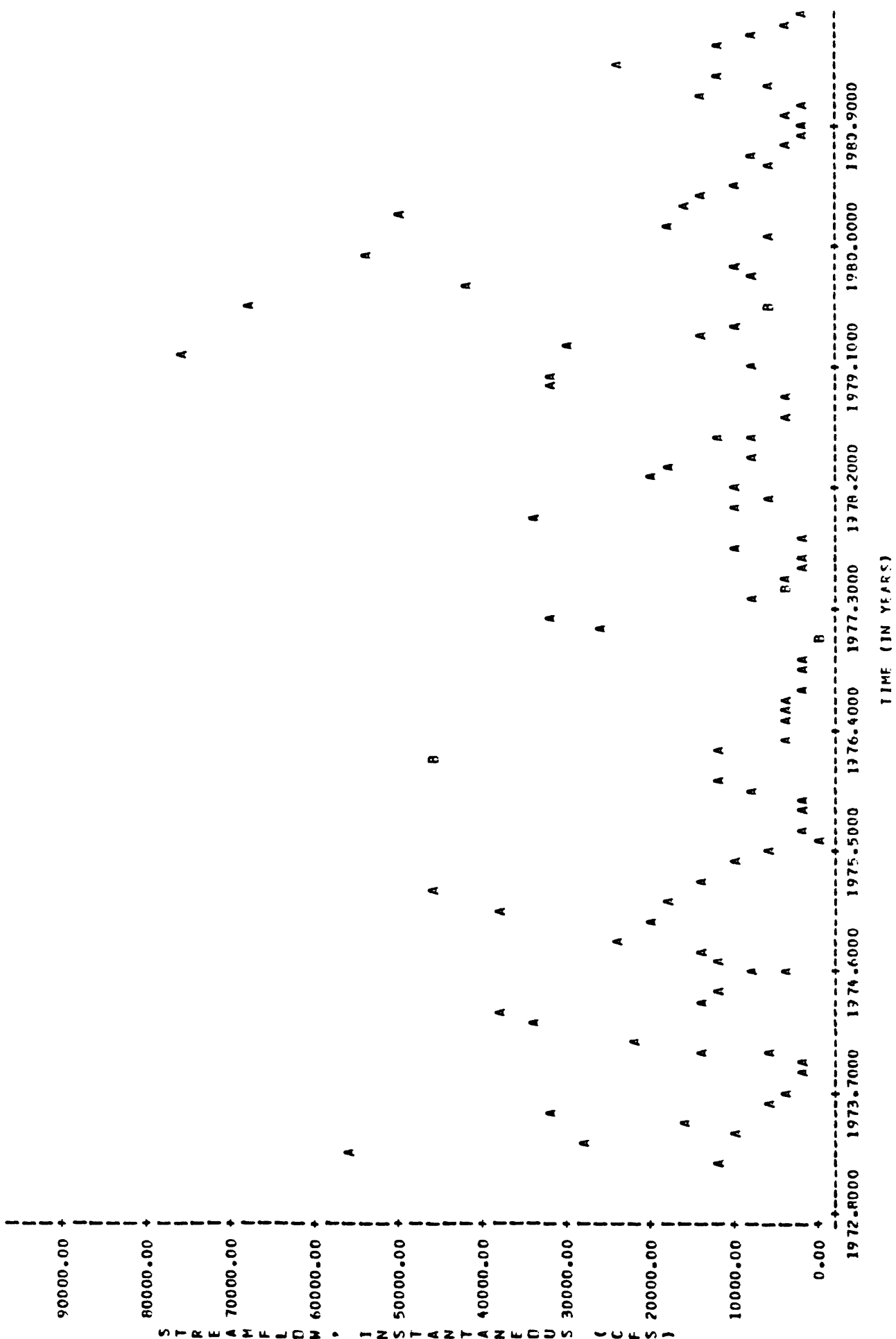
PROC PLOT produces line printer plots as shown in figure 5. More elaborate plots of SAS data can be made using PROC GPLOT (SAS Institute Inc., 1981b, p. 69). The procedure GPLOT is a part of the SAS/GRAPH system which is a computer graphics system for producing figures on terminal screens and plotters. A list of compatible graphics devices is given in the SAS/GRAPH User's Guide (SAS Institute, Inc., 1981b, p. 2).


```

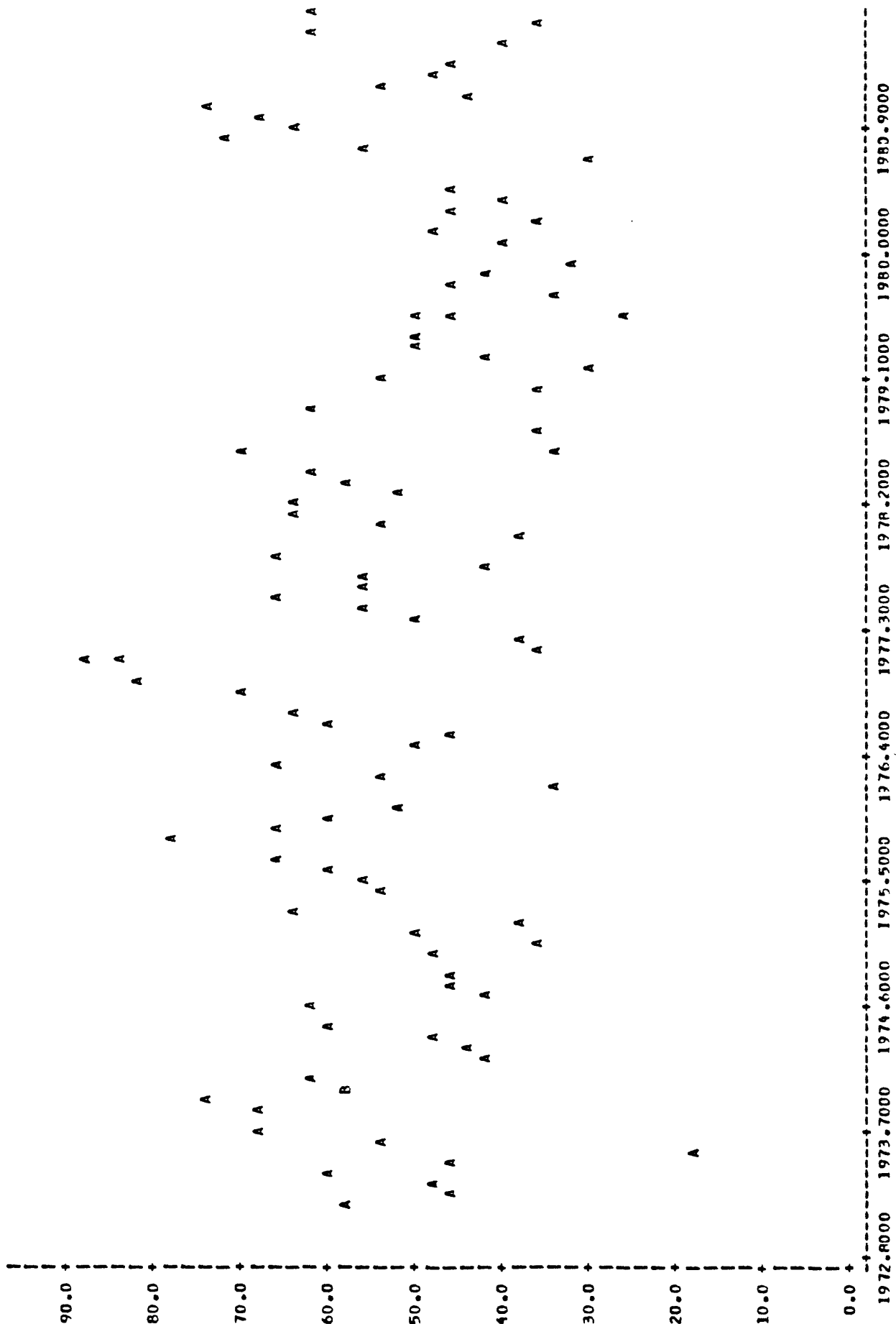
1 //F5 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD,PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //TREND DD DSN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 OPTIONS NOOVP;
8 DATA DATA;SET TREND,MONTHLY;
9 IF STATION=' 03374100 ';;
10 L00061=LOG(P00061);
11 KEEP STATION SNAME P00945 P00061 L00061 DECTIME;
12 PROC SORT DATA=DATA;BY STATION;
13 PROC REG DATA=DATA NOPRINT;BY STATION;
14 MODEL P00945=L00061;
15 OUTPUT OUT=PLOTDATA P=P R=FAC;
16 PROC SORT DATA=PLOTDATA;BY STATION;
17 PROC PLOT DATA=PLOTDATA;BY STATION;
18 PLOT P00061*DECTIME;
19 PLOT P00945*DECTIME;
20 PLOT FAC*DECTIME;
21 /*
22 //

```

Figure 5.--Example input and output of SAS statements to plot water-quality data and regression residuals as time series.



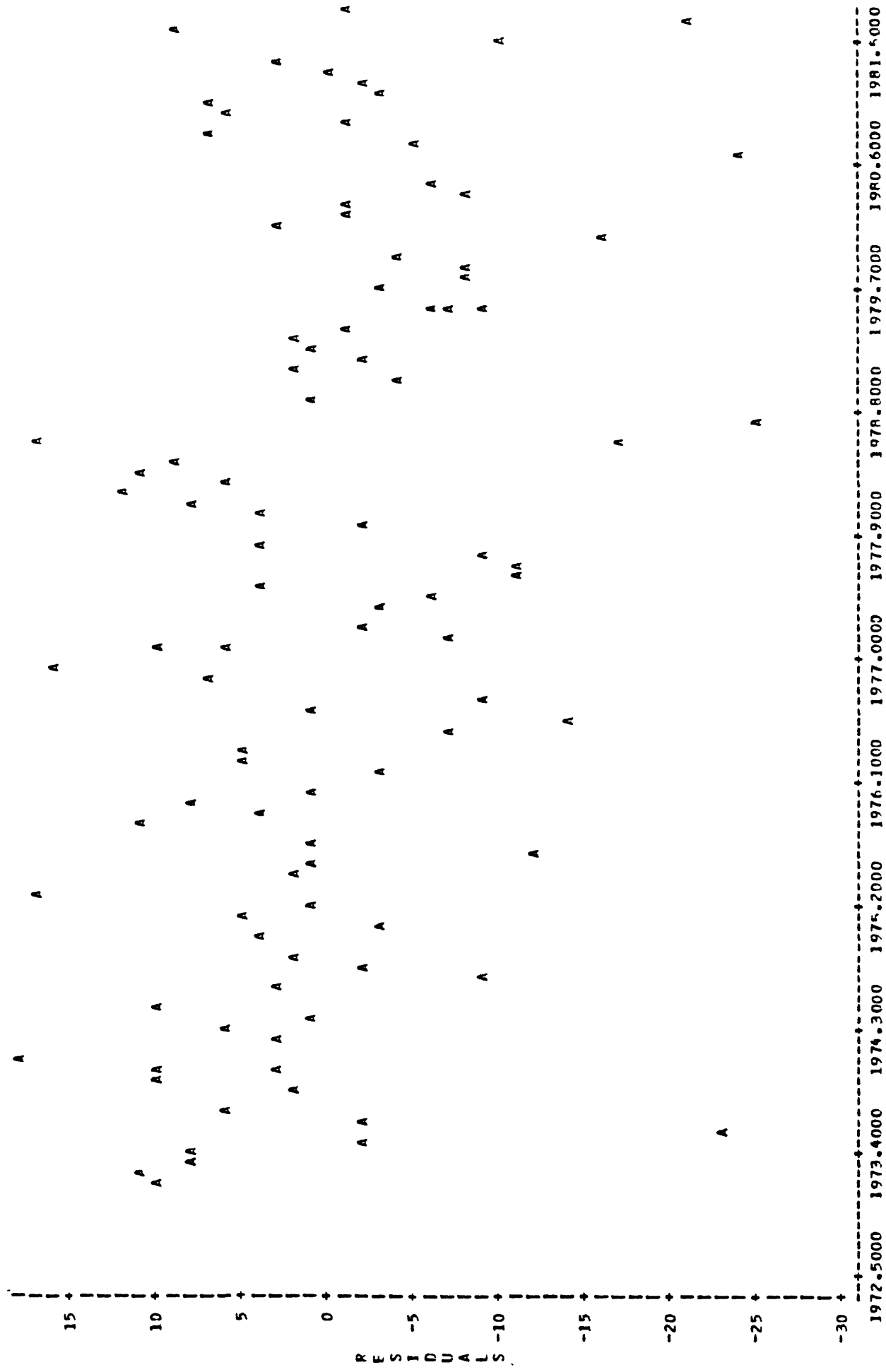
PLOT OF P00945*DECTIME LEGEND: A = 1 OBS, B = 2 OBS, ETC.



TIME (IN YEARS)

NOTE: 5 OBS HAD MISSING VALUES

PLOT OF FAC+DECTIME LEGEND: A = 1 ORS, R = 2 ORS, ETC.



TIME (IN YEARS)

NOTE: 5 ORS HAD MISSING VALUES

STATISTICAL PROCEDURES TO TEST FOR TRENDS IN WATER-QUALITY TIME SERIES

Two nonparametric tests for detecting trends in time series have been added to SAS at the U.S. Geological Survey Amdahl computer facility. PROC SEASKEN does the Seasonal Kendall test and slope estimator developed by Hirsch and others (1982). This procedure is suitable for detecting monotonic trends in time series with seasonality, missing values, or values reported as "less than." The procedure is not, however, robust against serial correlation. Additional information about the Seasonal Kendall test and slope estimator is given in the User's Guide presented in Appendix A.

PROC SEASRS tests for differences in the location parameters (mean, median, etc.) of two separate periods in a time series. The procedure uses a version of the Wilcoxon rank-sum test (Mann-Whitney test) modified to handle seasonality. This test is appropriate for detecting step trends (changes) in water quality before and after events such as construction of a dam or sewage treatment plant or an abrupt land-use change (construction, mining, clear cutting). Additional information about the modified Wilcoxon rank-sum test is given in the User's Guide presented in Appendix B.

In many cases, it may be appropriate to apply one of these PROC's to the concentration data as well as to the flow-adjusted concentration (FAC) data. Trends or changes in FAC may be interpreted as an indication that the stream-flow concentration relationship has changed; that is, for a given discharge one may expect different concentration values today versus some time in the past. In cases where there have been substantial changes in the flow frequency distribution (due to changing regulation, diversion, or consumption), it may be very useful to look for trends or changes in raw concentration data. This will be indicative of the combined effects of changes in human inputs to

the stream and changes in the flow frequency distribution on the frequency distribution of concentrations.

The presence of a large number of missing values or "less than" values (censoring) in the records being tested does not substantially affect the significance of the tests, although their power (ability to detect actual trends) may be reduced. PROC SEASKEN has been tested with as much as 50 percent of the seasonal values missing or 50 percent of them censored without any problem in significance.

The consequences of using these tests when data are serially correlated is that the actual significance of the test becomes higher than the nominal significance level. For monthly data, serial correlations are generally in a range such that actual significance may be about twice as high as the nominal. In other words, a p-level reported as 0.05 may more accurately be as high as 0.10. The problem becomes severe where the measurements are from a large body of water with a long residence time (by comparison with the sampling frequency). Aquifers, large lakes, or reservoirs, or the streams draining large lakes or reservoirs, may present problems. Keeping the parameters SEASON (see appendices A and B) small, typically no larger than 12, will help avoid serious problems due to serial correlation. When the correlation problem is thought to be severe, reducing SEASON (reducing the number of values per year used in the test) should prove to be helpful.

Figure 6 shows a program listing and output of a SAS job that uses PROC SEASKEN to test for time trends in the streamflow, sulfate concentration, and the flow-adjusted sulfate concentrations used in the regression example shown in figure 3. The flow-adjusted concentrations were calculated by the REG procedure using the LOGLIN model shown in figure 3 and are included in the output data set generated by PROC REG.

```

1 //F6 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSNT=NULLFILE,DSN2=NULLFILE
5 //TRENU DD USN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 DATA DATA;SET TRENU.MONTHLY;
8 IF STATION=' 03374100 ' ;
9 L00061=LOG(P00061);
10 KEEP STATION SNAME P00945 P00061 L00061 DECTIME;
11 PROC SORT DATA=DATA;BY STATION;
12 PROC REG DATA=DATA NOPRINT;BY STATION;
13 MODEL P00945=L00061;
14 OUTPUT OUT=PLOTDATA P=P R=FAC;
15 PROC SORT UATA=PLOTDATA;BY STATION;
16 PROC SEASKEN DATA=PLOTDATA SEASON=12;BY STATION;
17 VAR DECTIME P00061 P00945 FAC;
18 /*
19 //

```

Figure 6.---Example input and output of SAS statements to do the Seasonal Kendall test and slope estimator procedure on a water-quality data time series.

SEASONAL KENDALL TEST AND SLOPE ESTIMATOR FOR TREND MAGNITUDE

WATER YEARS 1973-1981

VARIABLE	N OBS	N VALS	TAU	P LEVEL	SLOPE
P00061	103	98	-0.020	0.837	-60.00
P00945	98	96	-0.182	0.032	-1.333
FAC	98	96	-0.279	0.001	-1.066

N OBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 N VALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

The results of the SEASKEN test show that no significant trend is evident in streamflow but that both the sulfate and flow-adjusted sulfate concentrations exhibited highly significant decreasing trends ($p\text{-level}(p) < 0.05$) during the period of record used in the test.

This trend can be further examined using PROC SEASRS. The data used in this example are from a station draining a watershed containing considerable surface coal mining. The regulations promulgated under the Surface Mine Control and Reclamation Act of 1977 (PL 95-87) were implemented toward the latter third of the period of record for this station (May 1979). Figure 7 shows a program listing and output of a SAS job that uses PROC SEASRS to determine if differences exist in streamflow, sulfate concentration, and the flow-adjusted sulfate concentration (using the LOGLIN model of figure 3) before and after this date. The results of the SEASRS test show that no significant step trend is evident in streamflow ($p\text{-level} > 0.05$) but that sulfate concentration and the flow-adjusted sulfate concentration exhibit a highly significant trend ($p\text{-level} < 0.05$).

The reader should be aware that both the Seasonal Kendall test and the Seasonal Wilcoxon rank-sum test are exploratory in nature. The fact that the Seasonal Wilcoxon rank-sum test indicates a difference between the periods in the time series prior to and following the implementation of the surface mine regulations does not necessarily imply causality, nor does it imply that it will continue into the future.

```

1 //F7 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRUSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //TREND DD DSN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 DATA DATA;SET TREND,MONTHLY;
8 IF STATION=03374100 *;
9 L00061=LOG(P00061);
10 KEEP STATION SNAME P00945 P00061 L00061 DECTIME;
11 PRJC SORT DATA=DATA;BY STATION;
12 PROC REG DATA=DATA NOPRINT;BY STATION;
13 MODEL P00945=L00061;
14 OUTPUT OUT=FACDATA P=P R=FAC;
15 PROC SORT DATA=FACDATA;BY STATION;
16 PROC SEASRS DATA=FACDATA SEASON=12 DATE=1979.33;BY STATION;
17 VAR DECTIME P00061 P00945 FAC;
18 /*
19 //

```

Figure 7.--Example input and output of SAS statements to do the Seasonal Mann-Whitney-Wilcoxon rank sum and slope estimator procedure on a water-quality data time series.

MANN-WHITNEY-WILCOXAN RANK SUM TEST FOR SEASONAL DATA

VARIABLE	FIRST SERIES (1973-1979)		CUTOFF DATE	SECOND SERIES (1979-1981)		P LEVEL	STEP
	NOBS	NVALS		NOBS	NVALS		
P00061	73	69	1979.33	30	29	0.507	1050.
P00945	69	68	1979.33	29	28	0.006	-7.000
FAC	69	68	1979.33	29	28	0.006	-5.350

NOBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

APPENDIX A: PROC SEASKEN user's guide with examples

This procedure tests for monotonic trends in time series using a modified form of Kendall's tau (Kendall, 1975) derived by Hirsch and others (1982).

The null hypothesis for this test is that the probability distribution of the random variable does not change over time. In the Kendall's tau test, all possible pairs of data values are compared. If a later value (in time) is higher, a plus is recorded, if the later value is lower, a minus is recorded. If no trend exists in the data, the probability of a later value being higher or lower than any previous value is 0.50. In this case, the number of pluses should approximately equal the number of minuses. If the number of pluses greatly exceeds the number of minuses, the values later in the series are more often higher than those earlier in the series, indicating an uptrend. If the number of minuses greatly exceeds the number of pluses, a downtrend is indicated. In this modified Kendall's tau, the problem of seasonality is avoided by comparing only observations from the same season of the year. Thus, for monthly data with seasonality, January data is compared only with January data, and so on.

Trend magnitude is determined using the seasonal Kendall slope estimator (Hirsch and others, 1982). The slope estimate is taken to be the median of the slopes of the ordered pairs of data values compared in the Seasonal Kendall test:

A more complete discussion of the Seasonal Kendall test and Seasonal Kendall slope estimator and its use are given in Smith and others (1982).

SPECIFICATIONS

The following statements may be used in the SEASKEN procedure:

```
PROC SEASKEN SEASON = n options;
```

```
BY variables;
```

```
VAR variables;
```

The PROC SEASKEN statement invokes the procedure. The VAR statement is used to specify the numeric variables for the analysis. The BY statement is optional. The observations must be in chronological order. This can be done by using PROC SORT and sorting by the variables in the BY statement followed by DECTIME. For example:

```
PROC SORT; BY variable DECTIME;
```

PROC SEASKEN statement

The PROC SEASKEN statement has the form:

```
PROC SEASKEN SEASON = n options;
```

SEASON = n gives the number of seasonal values per year. For example, SEASON = 12 would be used for data collected monthly and SEASON = 52 would be used for data collected weekly. SEASON = 1 may be used for annual summary data, such as average annual streamflow. Each water year will be divided into n equal length seasons. Note that for SEASON=12, for example, the 12 seasons will not correspond exactly to the 12 calendar months since months are not of equal length. For each season of each year, there may be no observations (a missing value will be assumed as the seasonal value), one observation (that value will be used as the seasonal value), or several (not more than 400) observations (the median of the values will be used as the seasonal value).

This statement is not optional and must be included.

The options that may be used in the PROC SEASKEN statement follow.

DATA = SASdataset gives the name of the SAS dataset to be used by PROC SEASKEN. If it is omitted, the most recently created SAS data set is used.

DL1 - DL15 = detection limit

gives the detection limit of the analytical procedures used to determine the constituent values in the time series. The number associated with each call of the detection limit option refers to the order of the constituents on the VAR statement. The use of the detection limit option is most important when several different analytical procedures with different degrees of sensitivity were used to determine values of the same constituent in the time series. In this case, the highest detection limit of all the procedures should be used.

The detection limit option sets all values less than or equal to the detection limit equal to one-half the value of the detection limit.

This option would not be used when the test is applied to Flow-Adjusted Concentrations (FAC).

VAR statement

VAR dectime variables;

The names of the variables to be tested for trends are listed in the VAR statement. The first variable must contain the decimal time values corresponding to the times the observations in the time series were made. Dectime is in the form of decimal time; for example, 12 noon, June 1, 1975, will be 1975.4178. Up to 15 additional numeric variables may be included. The total number of seasonal values (number of seasons times the number of water years) may not exceed 1200. The VAR statement is not optional and must be included.

BY statement

BY variables;

A BY statement may be used with PROC SEASKEN to obtain separate analyses on observations in groups defined by the BY variables. The input data set must be sorted in order of the BY variables.

DETAILS

Missing Values

Missing values may appear in a time series used in SEASKEN. The existence of missing values presents no theoretical problem for applying the seasonal Kendall test for trend. Comparisons of data pairs where one member is missing are not included in the calculation of the test statistic. (Internally, missing values are represented as -1E31.)

One or more seasons may be devoid of data; for example, one may use SEASON=12 and have data only for the summer. In this case, all other months will be set to missing values.

Formulas

The test statistic, S_i , is:

$$S_i = \sum_{k=1}^{n_i-1} \sum_{j=k+1}^{n_i} \text{sgn}(X_{ij} - X_{ik})$$

Where n_i are the number of annual values for season i ,

X_{ij} the seasonal value for season i and year j ,

X_{ik} the seasonal value for season i and year k ,

and

$$\text{sgn}(\theta) = \begin{cases} 1 & \text{if } \theta > 0 \\ 0 & \text{if } \theta = 0 \\ -1 & \text{if } \theta < 0 \end{cases}$$

The expected value of S_i is 0, $E [S_i] = 0$ and its variance is:

$$\text{Var} [S_i] = \frac{n_i (n_i-1)(2n_i+5) - \sum_{t_i} t_i (t_i-1)(2t_i+5)}{18}$$

where t_i is the extent of a given tie (number of X's involved in tie)
for season i ,

and \sum_{t_i} denotes the summation over all ties.

The composite statistic of the seasonal statistics, S_i , is S' :

$$E [S'] = \sum_{i=1}^{\text{season}} E [S_i] = 0$$

and the variance of S' is:

$$\text{Var} [S'] = \sum_{i=1}^{\text{season}} \text{Var} [S_i] = \sum_{i=1}^{\text{season}} \frac{n_i (n_i-1)(2n_i+5)}{18}$$

The normal approximation with a continuity correction of 1 (toward zero) is used to estimate $p = \text{Prob} [|S'| \geq s]$ (the probability that S' will depart from zero by the amount s or more). The standard normal deviate, Z , is calculated by:

$$Z = \begin{cases} \frac{S' - 1}{(\text{Var } S')^{1/2}} & \text{if } S' > 0 \\ 0 & \text{if } S' = 0 \\ \frac{S' + 1}{(\text{Var } S')^{1/2}} & \text{if } S' < 0 \end{cases}$$

The p-level of the test is determined from Z using the standard normal distribution. The p-level is the probability of obtaining a Z value that is as large or larger in absolute value than the one obtained if the null hypothesis were true (i.e., there was actually no trend). If one pre-selects a significance level α (the risk of rejecting the null hypothesis when it is

actually true), then one should reject whenever the p-level is less than α .

A typical value for α is 0.05.

The statistic τ (tau) is:

$$\tau = \sum_{i=1}^{\text{season}} \frac{S_i}{n_i(n_i-1)/2}$$

Note that τ may only take on values between -1 and +1. Negative values indicate downwards trend, positive values indicate upwards trend. It is a type of rank correlaton coefficient between the variable and time.

The seasonal Kendall slope estimator is the median value of

$$d_{ijk} = (X_{ij} - X_{ik}) / (j-k) \text{ for all } (X_{ij}, X_{ik}) \text{ pairs}$$

where d_{ijk} is the slope between seasonal values for season i , year j and season i , year k with $j > k$.

The slope estimate is valid only when all of the data are reported as above the limit of detection. If there are only a few "less thans" it can be viewed as a reasonable approximation.

Printed output

For each variable SEASKEN prints (see figures A1 and A2)

1. the variable name (VARIABLE)
2. the variable label, if any
3. the total number of non-missing observations in the input data (NOBS)
4. the number of seasonal values constructed from the observations (NVALS)
5. the statistic tau (τ) (TAU)
6. The significance probability (p-level) of the trend, two-sided (P LEVEL)
7. The estimate of trend magnitude, in units per year (SLOPE)

EXAMPLES

Example 1: Annual Mean Streamflow

In this example (figure A-1), PROC SEASKEN is called to test for a trend in the annual mean streamflow observed at a U.S. Geological Survey streamflow gaging station. The data was entered from card input (lines 9-19).

Example 2: NASQAN Data

In this example (figure A-2), PROC SEASKEN is called to test for trends in 10 water quality constituents at three U.S. Geological Survey NASQAN stations. The data was previously stored as a SAS data set on a direct access storage device. Note that the SAS data set, TREND.MONTHLY was sorted by STATION (line 6) so that the BY statement option (line 8) could be used with SEASKEN.

```

1 //AT JOB
2 // CLASS
3 //PROCLIB DU DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //SYSIN DD *
6 DATA SFLOW;
7 INPUT WYEAR AMQ @;LIST;
8 DECTIME=WYEAR-0.25;
9 LABEL AMQ=ANNUAL MEAN STREAMFLOW;
10 CARDS;
11 1925 8010 1926 4760 1927 10800 1928 10300 1929 4950 1930 10500 1931 2200
12 1932 4660 1933 5930 1934 5630 1935 3430 1936 5230 1937 6960 1938 7140
13 1939 7720 1940 3570 1941 2410 1942 3650 1943 5950 1944 6880 1945 4910
14 1946 7660 1947 3430 1948 6930 1949 8370 1950 10400 1951 9440 1952 8290
15 1953 6300 1954 2840 1955 5100 1956 5410 1957 3520 1958 8690 1959 10200
16 1960 6340 1961 4650 1962 8330 1963 3880 1964 2700 1965 5160 1966 4280
17 1967 6050 1968 7450 1969 7280 1970 6840 1971 7130 1972 4450 1973 11200
18 1974 9930 1975 7180 1976 8180 1977 3080 1978 6440 1979 6870 1980 6470
19 1981 5540
20 ;
21 PROC SEASKEN SEASON=1;
22 VAR DECTIME AMQ;
23 /*
24 //

```

Figure A1.--Example input and output using PROC SEASKEN to test for a trend in annual mean streamflow.

SEASONAL KENDALL TEST AND SLOPE ESTIMATOR FOR TREND MAGNITUDE

WATER YEARS 1925-1981

VARIABLE	NOBS	NVALS	TAU	P LEVEL	SLOPE
ANNUAL MEAN STREAMFLOW	57	57	0.009	0.670	9.839

NOBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

```

1 //A2 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //TREND DD DSN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 PROC SURT DATA=TREND.MONTHLY;BY STATION;
8 PROC SEASKEN DATA=TREND.MONTHLY SEASON=12 DL3=0.1 DL4=0.01 DL7=0.1 DL10=0.1;
9 BY STATION;
10 VAR DECTIME P70300 P00410 P00630 P00665 P00915 P00925 P00930 P00940 P00945 P0095
11 0;
12 /*
13 //

```

Figure A2.--Example input and output using PROC SEASKEN to test for trends in water-quality constituents.

SEASONAL KENDALL TEST AND SLOPE ESTIMATOR FOR TREND MAGNITUDE
 WATER YEARS 1975-1981

VARIABLE	NOBS	NVALS	TAU	P LEVEL	SLOPE
P00300	77	74	0.119	0.259	4.600
P00410	66	63	-0.072	0.565	-1.292
P00630	76	73	0.223	0.024	.1000
P00655	75	73	-0.063	0.559	-.1000E-29
P00915	77	74	-0.077	0.469	-.5000
P00925	77	74	0.031	0.794	.1000E-29
P00930	77	74	-0.021	0.877	-.1000E-29
P00940	77	74	0.093	0.376	.1833
P00945	77	74	-0.175	0.087	-.6458
P00950	77	74	-0.046	0.598	-.1000E-29

NOBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

SEASONAL KENDALL TEST AND SLOPE ESTIMATOR FOR TREND MAGNITUDE
 WATER YEARS 1973-1981

VARIABLE	N OBS	NVALS	TAU	P LEVEL	SLOPE
P70300 SOLIDS, RESIDUE AT 180 DEG. C DISSOLVED	98	96	-0.073	0.400	-2.000
P00410 ALKALINITY FIELD (MG/L AS CaCO3)	86	84	-0.214	0.022	-4.667
P00630 NITROGEN, NH2+NH03 TOTAL (MG/L AS N)	93	91	0.236	0.007	.6000E-01
P00655 PHOSPHORUS, TOTAL (MG/L AS P)	92	90	0.107	0.235	.5000E-02
P00915 CALCIUM DISSOLVED (MG/L AS Ca)	98	96	-0.005	0.323	-.6667
P00925 MAGNESIUM, DISSOLVED (MG/L AS MG)	98	96	-0.065	0.456	-.2000
P00930 SODIUM, DISSOLVED (MG/L AS Na)	98	96	0.117	0.168	.1667
P00940 CHLORIDE, DISSOLVED (MG/L AS CL)	98	96	0.220	0.009	.5000
P00945 SULFATE DISSOLVED (MG/L AS SO4)	98	96	-0.182	0.032	-1.333
P00950 FLUORIDE, DISSOLVED (MG/L AS F)	98	96	-0.126	0.078	-.1000E-29

N OBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

STATION IDENTIFICATION NUMBER=03376500

SEASONAL KENDALL TEST AND SLOPE ESTIMATOR FOR TREND MAGNITUDE

WATER YEARS 1975-1981

VARIABLE	NRBS	NVALS	TAU	P LEVEL	SLOPE
P70300	72	70	0.011	0.957	.5635
P00410	63	62	-0.061	0.643	-.9750
P00630	75	72	0.174	0.098	.7550E-01
P00655	75	72	0.130	0.222	.6333E-02
P00915	72	71	0.056	0.628	.7500
P00925	73	71	-0.006	1.000	-.1000E-29
P00930	73	71	0.050	0.664	.6000E-01
P00940	76	72	0.174	0.098	.6333
P00945	74	72	-0.147	0.167	-1.717
P00950	74	72	-0.027	0.800	-.1000E-29

NRBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

Appendix B: PROC SEASRS user's guide with examples

The SEASRS procedure tests for differences in the location parameters of two separate periods in a time series using a modified version of the Wilcoxon rank-sum test (Bradley, 1968, p. 105). This test is equivalent to the Mann-Whitney test described in Conover (1971, p. 224). The null hypothesis for this test is that two populations comprised by data from two separate periods in a time series are identical. This test assumes that sampling was random.

If the null hypothesis is true, no distinction can be made between the n observations in the first period and the m observations in the second period of the time series. In effect, all were taken from a common population. Therefore, each of the possible combinations of the $n+m$ observations taken from the common population were equally likely to have become the two samples actually collected. For each of these possible combinations exists a value of the test statistic, W . This statistic is the sum of the ranks of the n observations within the combined ($n+m$ observations) sample. (The smallest value in the combined sample receives a rank of 1, the next smallest a rank of 2, etc.). The null hypothesis is rejected if the value of the test statistic W differs from the expected value of W by a preselected value corresponding to a desired probability.

For a more thorough discussion of the Wilcoxon rank-sum test, and examples of its use, the reader is referred to Bradley (1968, p. 105) and Hollander and Wolfe (1973, p. 68).

SPECIFICATIONS

The following statements may be used in the SEASRS procedure:

```
PROC SEASRS SEASON = n DATE = x options;
```

```
BY variables;
```

```
VAR variables.
```

The PROC SEASRS statement invokes the procedure. The VAR statement is used to specify the numeric variables for the analysis. The BY statement is optional. The observations must be in chronological order. This can be done by using PROC SORT and sorting by the variables in the BY statement followed by DECTIME. For example:

```
PROC SORT; BY variable DECTIME;
```

PROC SEASRS statement

The PROC SEASRS statement has the form:

```
PROC SEASRS SEASON = n DATE = x options;
```

SEASON = n gives the number of seasonal values per year.

For example, SEASON = 12 would be used for data collected monthly and SEASON = 52 would be used for data collected weekly. SEASON = 1 may be used for annual summary data, such as average annual streamflow. Each water year will be divided into n equal length seasons. Note that for SEASON=12, for example, the 12 seasons will not correspond exactly to the 12 calendar months, since months are not of equal length. For each season of each year, there may be no observations (a missing value will be assumed as the seasonal value), one observation (that value will be used as the seasonal value), or several (not more than 400) observations (the median of the observations will be used as the seasonal value).

This statement is not optional and must be included.

DATE = xxxx gives the time separating the two periods of the time series. DATE is in the form of decimal time, for example, 12 noon, June 1, 1975, will be 1975.4178. If the two periods in the time series are widely separated, any date in the gap will suffice. An observation occurring at precisely time DATE is placed in the second period.

This statement is not optional and must be included.

The options that may be used in the PROC SEASRS statement follow.

DATA = SASdataset gives the name of the SAS dataset to be used by PROC SEASRS. If it is omitted, the most recently created SAS data set is used.

DL1 - DL15 = detection limit

gives the detection limit of the analytical procedures used to determine the constituent values in the time series. The number associated with each call of the detection limit option refers to the order of the constituents on the VAR statement. The use of the detection limit option is most important when several different analytical procedures with different degrees of sensitivity were used to determine values of the same constituent in the time series. In this case, the highest detection limit of all the procedures should be used.

The detection limit option sets all values less than or equal to the detection limit equal to one-half the value of the detection limit.

This option would not be used when the test is applied to Flow-Adjusted Concentration (FAC).

VAR statement

VAR dectime variables;

The names of the variables to be tested are listed in the VAR statement. The first variable must contain the decimal time values corresponding to the times the observations in the time series were made. Dectime is in the form of decimal time; for example, 12 noon, June 1, 1975, will be 1975.4178. Up to 15 additional numeric variables may be included. The total number of seasonal values (number of seasons times the sum of the number of water years in the first period plus the number of water years in the second period) may not exceed 1200. The VAR statement is not optional and must be included.

BY statement

BY variables;

A BY statement may be used with PROC SEASRS to obtain separate analyses on observations in groups defined by the BY variables. The input data set must be sorted in order of the BY variables.

DETAILS

Missing values

Missing values may be included in time series used in SEASRS. The existence of missing values presents no theoretical problem for applying the Mann-Whitney-Wilcoxon rank-sum test for seasonal data. Ranks of observations with missing values are not included in the calculation of the test statistic. (Internally, missing values are represented as -1E31.)

One or more seasons may be devoid of data; for example, one may use SEASON=12 and have data only for the summer. In this case all other months will be set to missing values.

Formulas

The test statistic, W_i , is:

$$W_i = \sum_{j=1}^{n_i} R_n$$

where n_i is the number of annual values for season i in the first period in the time series,

and R_n are the ranks of the seasonal values for season i in the first period of the time series.

The expected value of W_i is:

$$E [W_i] = [n_i (n_i + m_i + 1)] / 2$$

where m_i is the number of annual values for season i in the second period in the time series,

and its variance is:

$$\text{Var} [W_i] = [n_i m_i (n_i + m_i + 1)] / 12.$$

The composite statistic of the seasonal statistic, W_i , is W and is defined as:

$$W = \sum_{i=1}^{\text{season}} W_i.$$

The expectation of W is:

$$E [W] = \sum_{i=1}^{\text{season}} E [W_i]$$

and its variance is:

$$\text{Var} [W] = \sum_{i=1}^{\text{season}} [W_i].$$

The normal approximation is used to estimate $p = \text{Prob} [|W - E[W]| \geq w]$ (the probability that W will depart from its expected value by the amount w or more). The standard normal deviate, Z , is calculated by:

$$Z = \frac{W - E[W]}{(\text{Var}[W])^{1/2}} .$$

The p -level of the test is determined from Z using the standard normal distribution. The p -level is the probability of obtaining a Z value that is as large or larger in absolute value than the one obtained if the null hypothesis were true (i.e., there was actually no trend). If one pre-selects a significance level α (the risk of rejecting the null hypothesis when it is actually true), then one should reject whenever the p -level is less than α . A typical value for α is 0.05.

An estimate of the magnitude of the step trend is taken as the median of the difference between all pairs of seasonal values, one from each period but of the same season.

The step trend estimate is valid only when all of the data are reported as above the limit of detection. If there are only a few "less thans" it can be viewed as a reasonable approximation.

Printed Output

For each variable SEASRS prints (see figures B1 and B2):

1. the variable name (VARIABLE)
2. the variable label, if any
3. the number of original observations (NOBS) and number of constructed seasonal values (NVALS) in the first series
4. the CUTOFF DATE value
5. the number of original observations and number of constructed seasonal values in the second series
6. the significance probability of the difference, two-sided (P LEVEL)
7. the estimate of the step trend (STEP)

EXAMPLES

Example 1: Annual Mean Streamflow

In this example (figure B-1), PROC SEASRS is called to test for differences in annual mean streamflows observed at a U.S. Geological Survey streamflow gaging station before and after construction of a series of reservoirs in the drainage basin. The data was entered from card input (lines 9-19).

Example 2: NASQAN Data

In this example (figure B-2), PROC SEASRS is called to test for differences in 10 water-quality constituents at three U.S. Geological Survey NASQAN stations before and after implementation of the Surface Mining Control and Reclamation Act of 1977. The second and third stations are located in watersheds mined for coal. The first station is not and serves as a control. Note that the SAS data set, TREND.MONTHLY was sorted by STATION (line 6) so that the BY statement option (line 8) could be used with SEASRS.

```

1 //B1 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //SYSIN DD *
6 DATA SFLOW;
7 INPUT WYEAR AMQ @;LIST;
8 DECTIME=WYEAR-0.25;
9 LABEL AMQ=ANNUAL MEAN STREAMFLOW;
10 CARDS;
11 1925 8010 1926 4760 1927 10800 1928 10300 1929 4950 1930 10500 1931 2200
12 1932 4660 1933 5930 1934 5630 1935 3430 1936 5230 1937 6960 1938 7140
13 1939 7720 1940 3570 1941 2410 1942 3650 1943 5950 1944 6880 1945 4910
14 1946 7660 1947 3430 1948 6930 1949 8370 1950 10400 1951 9440 1952 8290
15 1953 6300 1954 2840 1955 5100 1956 5410 1957 3520 1958 8690 1959 10200
16 1960 6340 1961 4650 1962 8330 1963 3880 1964 2700 1965 5160 1966 4280
17 1967 6050 1968 7450 1969 7280 1970 6840 1971 7130 1972 4450 1973 11200
18 1974 9930 1975 7180 1976 8180 1977 3080 1978 6440 1979 6870 1980 6470
19 1981 5540
20 ;
21 PROC SEASRS SEASON=1 DATE=1963.75;
22 VAR DECTIME AMQ;
23 /*
24 //

```

Figure B1.--Example input and output using PROC SEASRS to test for a step in annual mean streamflow.

MANN-WHITNEY-WILCOXAN RANK SUM TEST FOR SEASONAL DATA

VARIABLE	FIRST SERIES (1925-1963) NOBS NVALS	CUTOFF DATE	SECOND SERIES (1964-1981) NOBS NVALS	P	LFVEL	STEP
AMQ ANNUAL MEAN STREAMFLOW	39 39	1963.75	18 10	0.770		250.0

NOBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

```

1 //B2 JOB
2 // CLASS
3 //PROCLIB DD DSN=WRD.PROCLIB,DISP=SHR
4 //A EXEC WRDSAS,DSN1=NULLFILE,DSN2=NULLFILE
5 //TREND DD DSN=USERID.FILENAME,DISP=OLD
6 //SYSIN DD *
7 PROC SORT DATA=TREND.MONTHLY;BY STATION;
8 PROC SEASRS DATA=TREND.MONTHLY SEASON=12 DATE=1979.4 DL3=0.1 DL4=0.01 DL7=0.1 DL
9 10=0.1;BY STATION;
10 VAR DECTIME P70300 P00410 P00630 P00915 P00925 P00930 P00940 P00945 P0095
11 0;
12 /*
13 //

```

Figure B2.--Example input and output using PROC SEASRS to test for step trends in water-quality constituents.

MANN-WHITNEY-WILCOXAN RANK SUM TEST FOR SEASONAL DATA

VARIABLE	FIRST SERIES (1975-1979)		CUTOFF DATE	SECOND SERIES (1979-1981)		P LEVEL	STEP
	N OBS	N VALS		N OBS	N VALS		
P70300	49	47	1979.40	28	27	0.526	9.000
P00410	49	47	1979.40	17	16	0.662	2.000
P00630	49	47	1979.40	27	26	0.009	.3000
P00665	49	47	1979.40	27	26	0.430	.1000E-01
P00915	49	47	1979.40	28	27	0.566	-1.000
P00925	49	47	1979.40	28	27	0.652	.0
P00930	49	47	1979.40	28	27	0.030	-1.100
P00940	49	47	1979.40	28	27	0.275	-1.000
P00945	49	47	1979.40	28	27	0.000	-5.000
P00950	49	47	1979.40	28	27	0.260	.0

N OBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 N VALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

STATION IDENTIFICATION NUMBER=03374100

MANN-WHITNEY-WILCOXAN RANK SUM TEST FOR SEASONAL DATA

VARIABLE	FIRST SERIES (1973-1979)		SECOND SERIES (1979-1981)		CUTOFF DATE	P LEVEL	STEP
	N OBS	N VALS	N OBS	N VALS			
P70300	70	69	28	27	1979.40	0.166	-14.50
P00410	70	69	16	15	1979.40	0.211	-14.00
P00630	65	64	28	27	1979.40	0.001	.5000
P00665	64	63	28	27	1979.40	0.654	.1000E-01
P00915	70	69	28	27	1979.40	0.578	-2.000
P00925	70	69	28	27	1979.40	0.245	-2.000
P00930	70	69	28	27	1979.40	0.578	.6000
P00940	70	69	28	27	1979.40	0.608	1.000
P00945	70	69	28	27	1979.40	0.006	-7.000
P00950	70	69	28	27	1979.40	0.746	.0

N OBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 N VALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

MANN-WHITNEY-WILCOXAN RANK SUM TEST FOR SEASONAL DATA

VARIABLE	FIRST SERIES (1975-1979)		CUTOFF DATE	SECOND SERIES (1979-1981)		P LEVEL	STEP
	N OBS	N VALS		N OBS	N VALS		
P70300	47	46	1979.40	25	24	0.788	-3.500
P00410	47	46	1979.40	16	16	0.279	-10.000
P00630	46	46	1979.40	29	26	0.168	.3000
P00665	46	46	1979.40	29	25	0.133	.3000E-01
P00915	45	45	1979.40	27	26	0.842	-1.000
P00925	46	45	1979.40	27	26	0.139	-2.000
P00930	46	45	1979.40	27	26	0.547	-1.000
P00940	47	46	1979.40	29	26	0.512	-1.000
P00945	47	46	1979.40	27	26	0.089	-8.000
P00950	47	46	1979.40	27	26	0.352	.0

N OBS IS THE NUMBER OF NON-MISSING OBSERVATIONS IN THE ORIGINAL DATA.
 N VALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES CONSTRUCTED.

Appendix C: Source code for SAS procedures

Two separate programs are required to add a procedure to the Statistical Analysis System at a user's installation, a parsing module and a procedure module. The parsing module acts as a control program for the procedure and defines permissible options and parameters for the procedure. The procedure module inputs the appropriate data and computes the test statistic. The parsing module for Proc SEASKEN is given in figure C-1. The procedure module is given in figure C-2. The parsing module for Proc SEASRS is given in figure C-3. The procedure module is given in figure C-4. Information about SAS parsing and procedure modules can be found in the SAS Programmer's Guide (SAS Institute, Inc., 1981a). Only minimal JCL is shown since JCL is highly installation dependent.

```

1 //C1 JOB
2 // CLASS
3 // EXEC ASMSAS
4 //ASM.SYSIN DD *
5 PRINT NOGEN
6 SASPROC NAME=SEASKEN,LOADMOD=SEASKEN2,DEFLIST=1,
7 DEFMODE=NUMERIC
8 PARS SASLIST DL1,1,DL2,2,DL3,3,DL4,4,DL5,5,MODE=NUMERIC
9 PARS SASLIST DL6,6,DL7,7,DL8,8,DL9,9,DL10,10,MODE=NUMERIC
10 PARS SASLIST DL11,11,DL12,12,DL13,13,DL14,14,MODE=NUMERIC
11 PARS SASLIST DL15,15,SEASON,16,MODE=NUMERIC
12 LISTS SASLIST VARIABLES,1,VAR,1,MODE=NUMERIC
13 SASEND
14 END
15 //LKED.SYSIN DD *
16 SETSSI AC000003
17 NAME SEASKEN(R)
18 /*
19 //

```

Figure C1.--Parsing module for the Seasonal Kendal1 test and slope estimator procedure.

```

1 //C2 JOB
2 // CLASS
3 //EXEC FORTSAS
4 //C.SYSIN DD *
5 COMMON /STATS/ NV,N,NOBS(15),NVALS(15),NDATES(2),TAU(15),
6 & ALPHA(15),SLOPE(15)
7 C NOTIFY SAS.
8 CALL SASFHX
9 C PROCESS THE DATA.
10 C GET NUMBER OF VARIABLES FROM SAS.
11 NV1=NOVAR(1)
12 C NV IS THE NUMBER OF VARIABLES EXCLUSIVE OF THE TIME VARIABLE.
13 NV=NV1-1
14 IF (NV.GT.0) GO TO 10
15 CALL VARERR
16 STOP
17 10 CONTINUE
18 C READ IN THE DATA.
19 CALL READ
20 C COMPUTE THE TEST STATISTICS.
21 CALL SEAS
22 C WRITE OUT RESULTS.
23 CALL OUTPUT
24 STOP
25 END
26 C SUBROUTINE TO READ IN THE OBSERVATIONS AND FORM THE SEASONAL VALUES.
27 SUBROUTINE READ
28 COMMON /STATS/ NV,N,NOBS(15),NVALS(15),NDATES(2),TAU(15),
29 & ALPHA(15),SLOPE(15)
30 COMMON /DATA/ X(1200,15)
31 COMMON /IUNIT/ IPRINT
32 LOGICAL*1 SETUP,HAVE(1200),EVEN
33 REAL*8 PARM,BASETM,PERYR,Y(16)
34 INTEGER L(15)
35 REAL SORT(400,15)
36 COMMON /LABCM/ NAME,LABEL
37 COMMON /NAMECM/ NTYPE,NPOS,NLNG,NVARO,NNAME,NLABEL,NFORM,NIFORM,
38 XNFL,NFD,NF,NJUST
39 INTEGER*2 NTYPE,NPOS,NLNG,NVARO,NFL,NFD,NF,NJUST
40 REAL*8 NNAME,NLABEL(5),NFORM,NIFORM
41 REAL*8 NAME(15),LABEL(5,15)
42 REAL XCHK/ -1.0E30/,XMISS/ -1.0E31/
43 M=0
44 KORS=0
45 C GET SPECIFICATION OF EACH VARIABLE.
46 DO 10 I=1,NV
47 C CALL SAS TO LOAD COMMON BLOCK.
48 CALL NAMEV(1,I+1,NTYPE)
49 C SAVE NAME.
50 NAME(I)=NNAME

```

Figure C2.--Procedure module for the Seasonal Kendal11 test and slope estimator procedure.


```

51 C SAVE LABEL.
52 DO 11 J=1,5
53 LABEL(J,I)=NLABEL(J)
54 11 CONTINUE
55 NOBS(I)=0
56 NVALS(I)=0
57 10 CONTINUE
58 SETUP=.FALSE.
59 CYCLE TO GET EACH OBSERVATION.
60 CALL INPUT(IEND)
61 *WRAP UP IF ALL OBSERVATIONS HAVE BEEN PROCESSED.
62 IF(IEND.EQ.1)GO TO 40
63 GET AN OBSERVATION.
64 K OBS=K OBS+1
65 CALL VARX(1,Y)
66 IF(SETUP) GO TO 80
67 FIRST OBSERVATION, GET READY.
68 CALCULATE TIME OF OCT. 1 OF WATER YEAR OF FIRST OBSERVATION.
69 YEAR=Y(1)+0.25
70 BASETM=FIX(YEAR)-0.25
71 NDATE(1)=YEAR
72 GET NUMBER OF SEASONS.
73 PERYR=PARM(16)
74 DO 70 I=1,1200
75 HAVE(I)=.FALSE.
76 70 CONTINUE
77 LAST=1
78 SETUP=.TRUE.
79 80 CONTINUE
80 C PROCESS OBSERVATION.
81 M=(Y(1)-BASETM)*PERYR+1.0
82 IF (M.GE.LAST) GO TO 200
83 WRITE (IPRINT,9001) K OBS
84 9001 FORMAT (' OBSERVATION NUMBER',I10,
85 & ' IS OUT OF CHRONOLOGICAL ORDER.')
86 STOP
87 200 CONTINUE
88 IF (M.LE.1200) GO TO 210
89 WRITE (IPRINT,9002) K OBS
90 9002 FORMAT (' OBSERVATION NUMBER',I10,' IS LATER THAN SEASON 1200.')
91 STOP
92 210 CONTINUE
93 C MULTIPLE OBSERVATIONS?
94 IF (HAVE(M)) GO TO 90
95 C NO. SAVE.
96 HAVE(M)=.TRUE.
97 K=2
98 LAST=M
99 DO 30 I=1,NV
100 C PREPARE FOR MULTIPLE OBSERVATIONS.

```

```

101 VAL=FIXMIS(Y(I+1))
102 L(I)=0
103 IF (VAL.GT.XCHECK) L(I)=1
104 IF (VAL.GT.XCHECK) NOBS(I)=NOBS(I)+1
105 IF (VAL.GT.XCHECK) NVALS(I)=NVALS(I)+1
106 X(M,I)=VAL
107 SORT(1,I)=VAL
108
109 30 CONTINUE
110 GO TO 20
111 C MULTIPLE OBSERVATIONS
112 90 CONTINUE
113 DO 110 I=1,NV
114 IF (K.EQ.2.AND.L(I).EQ.1) NVALS(I)=NVALS(I)-1
115 VAL=FIXMIS(Y(I+1))
116 IF (VAL.LT.XCHECK) GO TO 110
117 NOBS(I)=NOBS(I)+1
118 L(I)=L(I)+1
119 SORT(L(I),I)=VAL
120 110 CONTINUE
121 CALL INPUT(IEND)
122 C CYCLE UNTIL NEW SEASON.
123 IF (IEND.EQ.1) GO TO 120
124 KOB5=KOB5+1
125 CALL VARX(1,Y)
126 NEWM=(Y(1)-BASEM)*PERYR+1.0
127 IF (NEWM.NE.M) GO TO 120
128 K=K+1
129 C LIMIT TO 400 OBSERVATIONS, MISSING OR NOT, IN ONE SEASON.
130 IF (K.LE.400) GO TO 90
131 WRITE (IPRINT,9005) KOB5
132 9005 FORMAT (' AT OBSERVATION NUMBER,'I10,
133 & ', THERE WERE MORE THAN 400 OBSERVATIONS IN THAT SEASON. ')
134 STOP
135 C GET MEDIANS.
136 120 CONTINUE
137 DO 130 I=1,NV
138 C ANY NON-MISSING OBSERVATIONS?
139 IF (L(I).EQ.0) GO TO 130
140 YES.
141 NVALS(I)=NVALS(I)+1
142 IH=(L(I)+1)/2
143 EVEN=2*IH.EQ.L(I)
144 C FIND MEDIAN.
145 CALL VSRTA(SORT(1,I),L(I))
146 IF (.NOT.EVEN) X(M,I)= SORT(IH,I)
147 IF (EVEN) X(M,I)=(SORT(IH,I)+SORT(IH+1,I))/2.0
148 130 CONTINUE
149 LAST=M
150 C CONTINUE PROCESSING IF END-OF-FILE NOT ENCOUNTERED.
151 IF (IEND.NE.1) GO TO 80

```

```

151 40 IF (M.GT.1) GO TO 50
152 WRITE (IPRINT,9003)
153 9003 FORMAT (' THERE ARE FEWER THAN 2 SEASONS. ')
154 STOP
155 50 CONTINUE
156 N=LAST
157 NDATES(2)=Y(1)+0.25
158 IF (LAST*(LAST/PERYR+1.0).LE.60000.0) GO TO 60
159 WRITE (IPRINT,9004)
160 9004 FORMAT (' THE COMBINATION OF YEARS AND SEASONS EXCEEDS',
161 & ' AN INTERNAL LIMIT. ')
162 STOP
163 60 CONTINUE
164 DO 150 M=1, LAST
165 IF (HAVE(M)) GO TO 150
166 DO 140 I=1, NV
167 X(M,I)=XMISS
168 140 CONTINUE
169 150 CONTINUE
170 RETURN
171 END
172 C PERFORM KENDALL TEST.
173 SUBROUTINE SEAS
174 COMMON /STATS/ NV,N,NOBS(15),NVALS(15),NDATES(2),TAU(15),
175 & ALPHA(15),SLOPE(15)
176 COMMON /DATA/ X(1200,15)
177 REAL Y(60000)
178 REAL *8 PARM
179 REAL XCHEK/-1.0E30/
180 LOGICAL ODD
181 LOGICAL *1 WASTIE(1200)
182 NPER=PARM(16)
183 PERYR=NPER
184 DO 1 J=1,NV
185 C BYPASS DETECTION LIMIT PROCESSING IF NONE SPECIFIED.
186 IF (PARM(J).EQ.0.000) GO TO 50
187 DLIMIT=PARM(J)
188 DO 100 I=1,N
189 IF (X(I,J).LE.DLIMIT.AND.X(I,J).GT.0.0) X(I,J)=0.5*DLIMIT
190 100 CONTINUE
191 50 CONTINUE
192 C ZERO OUT THE COUNTERS.
193 DO 60 I=1,N
194 WASTIE(I)=.FALSE.
195 60 CONTINUE
196 NPLUS=0
197 NMINUS=0
198 NCOMPT=0
199 VARTOT=0.0
200 INDEX=0

```

```

201 FIXVAR=0.0
202 LOOP THROUGH THE SEASONS.
203 DO 10 ISEAS=1,NPER
204 NCOMP=0
205 N1=N-NPER
206 LOOP THROUGH YEARS FOR VALUES IN SEASON ISEAS.
207 DO 20 ISTART=ISEAS,N1,NPER
208 VALID VALUES?
209 IF (X(ISTART,J).LE.XCHEK) GO TO 20
210 VALUE IS ALWAYS TIED WITH ITSELF.
211 NTIE=1
212 N2=ISTART+NPER
213 TRY EACH LATER SEASON.
214 DO 30 IEND=N2,N,NPER
215 VALID VALUE?
216 IF (X(IEND,J).LE.XCHEK) GO TO 30
217 COMPARE.
218 NCOMP=NCOMP+1
219 INDEX=INDEX+1
220 YY=(X(IEND,J)-X(ISTART,J))/((IEND-ISTART)/PERYR)
221 IF (YY.GT.0.0) NPLUS=NPLUS+1
222 IF (YY.LT.0.0) NMINUS=NMINUS+1
223 IF (YY.EQ.0.0) NTIE=NTIE+1
224 MARK VALUES THAT ARE TIED.
225 IF (YY.EQ.0.0) WASTIE(IEND)=.TRUE.
226 SAVE ADJUSTED DIFFERENCES.
227 Y(INDEX)=YY
228 30 CONTINUE
229 C UPDATE VARIANCE CORRECTION IF TIES OCCURED AND TIES
230 WERE NOT COUNTED BEFORE.
231 IF (NTIE.NE.1.AND..NOT.WASTIE(ISTART)) FIXVAR=FIXVAR+NTIE*(NTIE-
232 & 1.0)*(2.0+NTIE+5.0)/18.0
233 20 CONTINUE
234 C ACCUMULATE THIS MONTH'S RESULTS.
235 NCOMPT=NCOMPT+NCOMP
236 NMONTH=(1.0+SQRT(1.0+8.0*NCOMP))/2.0
237 VARTOT=VARTOT+NMONTH*(NMONTH-1.0)*(2.0+NMONTH+5.0)/18.0
238 10 CONTINUE
239 C DONE COMPARING.
240 S=NPLUS-NMINUS
241 VARTOT=VARTOT-FIXVAR
242 WERE THERE ANY VALID COMPARISONS AND AT LEAST TWO DIFFERENT VALUES?
243 IF (NCOMPT.GT.0.AND.VARTOT.GT.0.0) GO TO 40
244 NONE. GO HOME EMPTY.
245 TAU(J)=0.0
246 ALPHA(J)=1.0
247 SLOPE(J)=0.0
248 GO TO 1
249 40 CONTINUE
250 C CALCULATE THE STATISTICS.

```

```

251 TAU(J)=S/NCOMPT
252 CONTINUITY CORRECTION.
253 IF (S.GT.0.0) S=S-1
254 IF (S.LT.0.0) S=S+1
255 Z=S/SQRT(VARTOT)
256 COMPARE TO THE STANDARD NORMAL DISTRIBUTION.
257 IF (Z.LE.0.0) ALPHA(J)=2.0*CDFN(Z)
258 IF (Z.GT.0.0) ALPHA(J)=2.0*(1.0-CDFN(Z))
259 SORT THE DIFFERENCES. VSRTA IS AN IMSL ROUTINE.
260 CALL VSRTA(Y,INDEX)
261 PICK THE MEDIAN.
262 ODD=MOD(INDEX,2).EQ.1
263 IF (ODD) YMED=Y((INDEX+1)/2)
264 IF (.NOT.ODD) YMED=0.5*(Y(INDEX/2)+Y((INDEX/2)+1))
265 SLOPE(J)=YMED
266 IF (SLOPE(J).NE.0.0) GO TO 1
267 ADJUST FOR THE FACT THAT TAU AND ALPHA MAY SAY THERE IS A
268 SIGNIFICANT TREND BUT THE ESTIMATE OF THE SLOPE IS
269 ZERO DUE TO TIES.
270 IF (NMINUS.GT.NPLUS) SLOPE( J )=-1.0E-30
271 IF (NMINUS.LT.NPLUS) SLOPE( J )=1.0E-30
272 1 CONTINUE
273 RETURN
274 END
275 C REPORT RESULTS.
276 SUBROUTINE OUTPUT
277 COMMON /LABCM/ NAME,LABEL
278 REAL*8 NAME(15),LABEL(5,15)
279 COMMON /STATS/ NV,N,NOBS(15),NVALS(15),NDATES(2),TAU(15),
280 & ALPHA(15),SLOPE(15)
281 COMMON /IOUNITY/ IPRINT
282 STANDARD SAS HEADING.
283 CALL STITLE(0,LINES)
284 WRITE (IPRINT,6001) NDATES
285 FORMAT (/26X,'SEASONAL KENDALL TEST AND SLOPE ESTIMATOR FOR',
286 & ' TREND MAGNITUDE'//40X,'WATER YEARS ',I4,'-',I4//)
287 WRITE (IPRINT,6002)
288 FORMAT (' VARIABLE',50X,'NOBS NVALS TAU P',
289 & ' LEVEL SLOPE'//1X,107('-'//)
290 DO 10 I=1,NV
291 WRITE (IPRINT,6003) NAME(I),(LABEL(J,I),J=1,5),NOBS(I),NVALS(I),
292 & TAU(I),ALPHA(I),SLOPE(I)
293 FORMAT (1X,A8,3X,5A8,2I10,2F10.3,G15.4//)
294 10 CONTINUE
295 WRITE (IPRINT,6004)
296 FORMAT (',', NOBS IS THE NUMBER OF NON-MISSING OBSERVATIONS',
297 & ', IN THE ORIGINAL DATA.',
298 & ', NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES',
299 & ', CONSTRUCTED.')
300 RETURN

```

```

301 END
302 CUMULATIVE DISTRIBUTION FUNCTION FOR THE NORMAL DISTRIBUTION.
303 C PRIMARY REFERENCE IS ABRAMOWITZ AND STEGUN,
304 C NBS HANDBOOK OF MATHEMATICAL FUNCTIONS, EQ. 26.2.19.
305 FUNCTION CDFN(X)
306 IF (X) 10,20,30
307 10 CONTINUE
308 IF (X.LT.-6.0) GO TO 40
309 T=-X
310 CDFN=0.5/(1.0+0.0498673470*T+0.0211410061*T**2+0.0032776263*T**3
311 X+0.380036E-4*T**4+0.488906E-4*T**5+0.53830E-5*T**6)**16
312 RETURN
313 20 CONTINUE
314 CDFN=0.5000
315 RETURN
316 30 CONTINUE
317 IF (X.GT.6.0) GO TO 50
318 CDFN=1.0-0.5/(1.0+0.0498673470*X+0.0211410061*X**2+0.0032776263*X
319 X**3+0.380036E-4*X**4+0.488906E-4*X**5+0.53830E-5*X**6)**16
320 RETURN
321 40 CONTINUE
322 CDFN=0.0000
323 RETURN
324 50 CONTINUE
325 CDFN=1.0000
326 RETURN
327 END
328 C CONVERT A SAS MISSING VALUE TO ONE RECOGNIZED BY THESE ROUTINES.
329 FUNCTION FIXMIS(X)
330 REAL*8 X,VALUE
331 EQUIVALENCE(VALUE,MISX)
332 VALUE=X
333 IF (VALUE.EQ.0.0DD.AND.MISSX.NE.0) VALUE=-1.0D31
334 FIXMIS=VALUE
335 RETURN
336 END
337 //LKED.SYSIN DD *
338 ENTRY ENTRY THIS STATEMENT MUST BE PRESENT
339 SETSSI AD00000
340 NAME SEASKEN2(R)
341 /*
342 //

```

```

1 //C3 JOB
2 // CLASS
3 // EXEC ASMSAS
4 //ASM.SYSIN DD *
5 PRINT NOGEN
6 SASPROC NAME=SEASRS,LOADMOD=SEASRS2,DEFLIST=1,
7 DEFMODE=NUMERIC
8 PARMS DL1,1,DL2,2,DL3,3,DL4,4,DL5,5,MODE=NUMERIC
9 PARMS DL6,6,DL7,7,DL8,8,DL9,9,DL10,10,MODE=NUMERIC
10 PARMS DL11,11,DL12,12,DL13,13,DL14,14,MODE=NUMERIC
11 PARMS DL15,15,SEASON,16,DATE,17,MODE=NUMERIC
12 LISTS SASLIST VARIABLES,1,VAR,1,MODE=NUMERIC
13 SASEND
14 END
15 //LKED.SYSIN DD *
16 SETSSI ACU00003
17 NAME SEASRS(R)
18 /*
19 //

```

Figure C3.--Parsing module for the Seasonal Mann-Whitney-Wilcoxon rank sum test and slope estimator procedure.

```

1 //C4 JOB
2 // CLASS
3 // EXEC FORTSAS
4 //C.SYSIN DD *
5 C NOTIFY SAS.
6 CALL SASFHX
7 C PROCESS THE DATA.
8 C GET NUMBER OF VARIABLES FROM SAS.
9 NV1=NOVAR(1)
10 C NV IS THE NUMBER OF VARIABLES EXCLUSIVE OF THE TIME VARIABLE.
11 NV=NV1-1
12 IF (NV.GT.0) GO TO 10
13 CALL VARERR
14 STOP
15 10 CONTINUE
16 C READ IN DATA.
17 CALL READ (NV1)
18 DO 20 J=1,NV
19 C COMPUTE RANK SUM FOR EACH VARIABLE.
20 CALL SEASRS (J)
21 20 CONTINUE
22 C WRITE OUT RESULTS.
23 CALL OUTPUT (NV)
24 STOP
25 END
26 C SUBROUTINE TO READ IN THE OBSERVATIONS AND FORM THE SEASONAL VALUES.
27 SUBROUTINE READ (NV1)
28 COMMON /DATA/ X(1200,15),NX1,NX2,ALPHA(15),DF(15),CUTOFF,
29 & NOBS1(15),NOBS2(15),NVALS1(15),NVALS2(15),NDATES(4)
30 REAL*8 CUTOFF
31 COMMON /LABINF/ NAME(15),LABEL(5,15)
32 INTEGER NOBS(15),NVALS(15)
33 DIMENSION L(15)
34 COMMON /IOUNIT/ IPRINT
35 REAL*8 NAME,LABEL,Y(16),PARM,BASETM,PERYR
36 REAL SORT(400,15)
37 LOGICAL*1 HAVE(1200),SETUP,SECOND,EVEN
38 COMMON /NAMECM/ NTYPE,NPOS,NLNG,NVARO,NNAME,NLABEL,NFORM,NIFORM,
39 & NFL,NFD,NF,NJUST
40 INTEGER*2 NTYPE,NPOS,NLNG,NVARO,NFL,NFD,NF,NJUST
41 REAL*8 NNAME,NLABEL(5),NFORM,NIFORM
42 REAL XCHK/ -1.0E30/,XMISS/ -1.0E31/
43 M=0
44 KOB5=0
45 NV=NV1-1
46 C GET SPECIFICATION OF EACH VARIABLE.
47 DO 10 I=1,NV
48 C CALL SAS TO LOAD COMMON BLOCK.
49 CALL NAMEV(1,I+1,NTYPE)
50 C SAVE NAME.

```

Figure C4.--Procedure module for the Seasonal Mann-Whitney-Wilcoxon rank sum test and slope estimator procedure.


```

51 NAME(I)=MNAME
52 SAVE LABEL.
53 DO 11 J=1,5
54 LABEL(J,I)=NLABEL(J)
55 11 CONTINUE
56 NOBS(I)=0
57 NVALS(I)=0
58 10 CONTINUE
59 SETUP=.FALSE.
60 SECOND=.FALSE.
61 MBASE=1
62 CYCLE TO GET EACH OBSERVATION.
63 CALL INPUT(IEND)
64 C WRAP UP IF ALL OBSERVATIONS HAVE BEEN PROCESSED.
65 IF (IEND.EQ.1) GO TO 40
66 C GET AN OBSERVATION.
67 K OBS=K OBS+1
68 CALL VARX(1,Y)
69 IF (SETUP) GO TO 70
70 C FIRST OBSERVATION, GET READY.
71 C CALCULATE TIME OF OCT. 1 OF WATER YEAR OF FIRST OBSERVATION.
72 YEAR=Y(1)+0.25
73 NDATES(1)=YEAR
74 BASET=FIX(YEAR)-0.25
75 C GET NUMBER OF SEASONS.
76 PERYR=PARM(16)
77 C GET CUTOFF DATE.
78 CUTOFF=PARM(17)
79 DO 60 I=1,1200
80 HAVE(I)=.FALSE.
81 60 CONTINUE
82 LAST=1
83 SETUP=.TRUE.
84 C PROCESS OBSERVATION.
85 70 CONTINUE
86 IF (SECOND.OR.Y(1).LT.CUTOFF) GO TO 80
87 C FIRST OBSERVATION AFTER CUTOFF DATE.
88 NDATES(2)=YEAR
89 SECOND=.TRUE.
90 MBASE=LAST+1
91 NX1=LAST
92 DO 75 I=1,NV
93 NOBS1(I)=NOBS(I)
94 NOBS(I)=0
95 NVALS1(I)=NVALS(I)
96 NVALS(I)=0
97 75 CONTINUE
98 YEAR=Y(1)+0.25
99 BASET=FIX(YEAR)-0.25
100 NDATES(3)=YEAR

```

```

101 80 CONTINUE
102   YEAR=Y(1)+0.25
103   M=(Y(1)-BASETM)*PERYR+MBASE
104   IF (M.GE.LAST) GO TO 90
105   WRITE (IPRINT,9001) K OBS
106   9001 FORMAT (' OBSERVATION NUMBER',I10,
107   & ' IS OUT OF CHRONOLOGICAL ORDER.')
108   STOP
109
110 90 CONTINUE
111   IF (M.LE.1200) GO TO 100
112   WRITE (IPRINT,9002) K OBS
113   9002 FORMAT (' OBSERVATION NUMBER',I10,' IS LATER THAN SEASON 1200.')
114   STOP
115 100 CONTINUE
116  C   MULTIPLE OBSERVATIONS?
117   IF (HAVE(M)) GO TO 110
118  C   NO. SAVE.
119   HAVE(M)=.TRUE.
120   K=2
121   LAST=M
122   DO 30 I=1,NV
123   C   PREPARE FOR POSSIBLE MULTIPLE OBSERVATIONS.
124   VAL=FIXMIS(Y(I+1))
125   L(I)=0
126   IF (VAL.GT.XCHECK) L(I)=1
127   IF (VAL.GT.XCHECK) NOBS(I)=NOBS(I)+1
128   IF (VAL.GT.XCHECK) NVALS(I)=NVALS(I)+1
129   SORT(I,I)=VAL
130   X(M,I)=VAL
131   GO TO 20
132  C   MULTIPLE OBSERVATIONS
133 110 CONTINUE
134   DO 130 I=1,NV
135   IF (K.EQ.2.AND.L(I).EQ.1) NVALS(I)=NVALS(I)-1
136   VAL=FIXMIS(Y(I+1))
137   IF (VAL.LT.XCHECK) GO TO 130
138   NOBS(I)=NOBS(I)+1
139   L(I)=L(I)+1
140   SORT(L(I),I)=VAL
141 130 CONTINUE
142   CALL INPUT(IEND)
143   CYCLE UNTIL NEW SEASON.
144   IF (IEND.EQ.1) GO TO 140
145   K OBS=K OBS+1
146   CALL VARX(1,Y)
147   NEWM=Y(1)-BASETM)*PERYR+MBASE
148   IF (NEWM.NE.M) GO TO 140
149   K=K+1
150  C   LIMIT TO 400 OBSERVATION, MISSING OR NOT, IN ONE SEASON.
      IF (K.LE.400) GO TO 110

```

```

151 WRITE (IPRINT,9005) K0BS
152 FORMAT (' AT OBSERVATION NUMBER',I10,
153 & ' THERE WERE MORE THAN 400 OBSERVATIONS IN THAT SEASON.')
154 STOP
155 GET MEDIANS
156 C
157 140 CONTINUE
158 DO 150 I=1,NV
159 C ANY NON-MISSING OBSERVATIONS?
160 IF (L(I).EQ.0) GO TO 150
161 C YES.
162 NVALS(I)=NVALS(I)+1
163 IH=(L(I)+1)/2
164 EVEN=2*IH.EQ.L(I)
165 C FIND MEDIAN.
166 CALL VSRTA(SORT(1,I),L(I))
167 IF (.NOT.EVEN) X(M,I)= SORT(IH,I)
168 IF (EVEN) X(M,I)=(SORT(IH,I)+SORT(IH+1,I))/2.0
169 150 CONTINUE
170 LAST=M
171 C CONTINUE PROCESSING IF END-OF-FILE NOT ENCOUNTERED.
172 IF (IEND.NE.1) GO TO 70
173 40 IF (M.GT.1) GO TO 50
174 WRITE (IPRINT,9003)
175 9003 FORMAT (' THERE ARE FEWER THAN 2 SEASONS.')
176 STOP
177 50 CONTINUE
178 IF (SECOND) GO TO 160
179 IF (IEND.NE.1) GO TO 70
180 WRITE (IPRINT,9004)
181 9004 FORMAT (' THERE ARE NO OBSERVATIONS AFTER THE CUTOFF DATE.')
182 STOP
183 160 CONTINUE
184 NDATES(4)=Y(1)+0.25
185 NX2=LAST-NX1
186 DO 170 I=1,NV
187 NOBS2(I)=NOBS(I)
188 NVALS2(I)=NVALS(I)
189 170 CONTINUE
190 C SET UNUSED CELLS TO MISSING.
191 DO 210 M=1,LAST
192 IF (HAVE(M)) GO TO 210
193 DO 200 I=1,NV
194 X(M,I)=XMISS
195 200 CONTINUE
196 210 CONTINUE
197 RETURN
198 END
199 C PERFORM RANK SUM TEST.
200 SUBROUTINE SEASRS (J)
COMMON /DATA/ X(1200,15),NX1,NX2,ALPHA(15),DF(15),CUTOFF,
& NOBS1(15),NOBS2(15),NVALS1(15),NVALS2(15),NDATES(4)

```

```

201 REAL*8 CUTOFF
202 DIMENSION Y(1200),R(1200)
203 REAL*8 PARM
204 REAL XCHEK/-1.0E30/
205 IF (NVALS1(J).EQ.0.OR.NVALS2(J).EQ.0) GO TO 700
206 NPER=PARM(16)
207 WS=U,U
208 EW=U,U
209 VW=U,U
210 LAST=NX1+NX2
211 C BYPASS DETECTION LIMIT PROCESSING IF NONE SPECIFIED.
212 IF (PARM(J).EQ.0.000) GO TO 20
213 DLIMIT=PARM(J)
214 DO 10 I=1, LAST
215 IF ( X(I,J).LE.DLIMIT.AND.X(I,J).GT.0.0) X(I,J)=0.5*DLIMIT
216 10 CONTINUE
217 20 CONTINUE
218 C LOOP THROUGH THE SEASONS.
219 DO 500 IM=1,NPER
220 C LOOP THROUGH YEARS FOR VALUES IN SEASON IM.
221 C FIRST PERIOD.
222 NI = 0
223 DO 60 I=IM,NX1,NPER
224 XX=X(I,J)
225 IF(XX.LT.XCHEK) GO TO 60
226 NI = NI + 1
227 Y(NI) = XX
228 60 CONTINUE
229 C SKIP SEASONS IF NO VALUES.
230 IF(NI.EQ.0) GO TO 500
231 C SECOND PERIOD.
232 MI = 0
233 DO 70 I=IM,NX2,NPER
234 XX=X(I+NX1,J)
235 IF(XX.LT.XCHEK) GO TO 70
236 MI = MI + 1
237 Y(MI+NI) = XX
238 70 CONTINUE
239 C SKIP SEASON IF NO VALUES.
240 IF (MI.EQ.0) GO TO 500
241 NN = MI + NI
242 CALL RANK(Y,R,NN)
243 RSUM = 0.0
244 DO 130 I = 1, NI
245 RSUM = RSUM + R(I)
246 130 CONTINUE
247 RSSQ=0.0
248 DO 140 I=1,NN
249 RSSQ=RSSQ+R(I)**2
250 140 CONTINUE

```

```

251 C ACCUMULATE STATISTICS.
252 WS=WS+RSUM
253 EW=EW+NI*(NN+1)/2.0
254 VM=VM+MI*NI*(RSSQ/NN-(NN+1)**2/4.0)/(NN-1)
255 500 CONTINUE
256 C CLACULATE FINAL STATISTICS.
257 IF(VM.LE.0.0) GO TO 700
258 Z=(-(WS-EW)/SQRT(VM))
259 IF (Z.LE.0.0) ALPHA(J)=2.0*CDFN(Z)
260 IF (Z.GT.0.0) ALPHA(J)=2.0*(1.0-CDFN(Z))
261 GET ESTIMATE OF JUMP.
262 CALL SEARSD(X(1,J),NX1,X(NX1+1,J),NX2,NPER,DF(J))
263 RETURN
264 700 CONTINUE
265 ALPHA(J)=1.0
266 DF(J)=0.0
267 RETURN
268 END
269 C CALCULATE AN ESTIMATE OF THE STEP TREND.
270 SUBROUTINE SEARSD(X1,NX1,X2,NX2,NPER,D)
271 COMMON /IUNIT/ IPRINT
272 REAL X1(NX1),X2(NX2)
273 REAL Y(12000)
274 REAL XCHECK/-.0E30/
275 LOGICAL EVEN
276 NY1=((NX1-1)/NPER)+1
277 NY2=((NX2-1)/NPER)+1
278 NMAX=NY1*NY2*12
279 IF (NMAX.LE.12000) GO TO 10
280 WRITE (IPRINT,9001) NX1,NX2
281 9001 FORMAT (' THE Y ARRAY IN SEARSD IS TOO SMALL. '/
282 & 'NX1=',I6,'NX2=',I6)
283 STOP
284 10 CONTINUE
285 C CALCULATE THE DIFFERENCE BETWEEN EACH PAIR OF VALUES - ONE IN
286 C THE FIRST SERIES, THE OTHER OF THE SECOND, AND BOTH OF THE
287 C SAME SEASON.
288 K=0
289 DO 40 IM=1,NPER
290 DO 30 I=IM,NX1,NPER
291 XF=X1(I)
292 IF (XF.LT.XCHECK) GO TO 30
293 DO 20 J=IM,NX2,NPER
294 XL=X2(J)
295 IF (XL.LT.XCHECK) GO TO 20
296 K=K+1
297 Y(K)=XL-XF
298 20 CONTINUE
299 30 CONTINUE
300 40 CONTINUE

```

```

301 GET MEDIAN.
302 CALL VSRTA(Y,K)
303 IH=(K+1)/2
304 EVEN=2*IH.EQ.K
305 IF (EVEN) D=(Y(IH)+Y(IH+1))/2.0
306 IF (.NOT.EVEN) D=Y(IH)
307 RETURN
308 END
309
310 REPORT RESULTS.
311 SUBROUTINE OUTPUT (NV)
312 COMMON /DATA/ X(1200,15),NX1,NX2,ALPHA(15),DF(15),CUTOFF,
313 & NOBS1(15),NOBS2(15),NVALS1(15),NVALS2(15),NDATES(4)
314 REAL*8 CUTOFF
315 COMMON /LABINF/ NAME(15),LABEL(5,15)
316 REAL*8 NAME,LABEL
317 COMMON /IOUNTIT/ IPRINT
318 STANDARD SAS TITLE.
319 CALL STITLE (0,LINES)
320 WRITE (IPRINT,6001)
321 6001 FORMAT (/59X,'MANN-WHITNEY-WILCOXAN RANK SUM TEST FOR SEASONAL',
322 & ' DATA'////)
323 WRITE (IPRINT,6002) NDATES
324 6002 FORMAT (53X,' FIRST SERIES ',16X,' SECOND SERIES ' /
325 & 52X,' (' ,I4,'-',I4,') ',16X,' (' ,I4,'-',I4,')' /
326 & ' VARIABLE', NOBS NVALS ', ' CUTOFF DATE ',
327 & ' NOBS NVALS',
328 & 5X,' P LEVEL',5X,' STEP',1X,123('-'//)
329 DO 10 I=1,NV
330 WRITE (IPRINT,6003) NAME(I),(LABEL(J,I),J=1,5),NOBS1(I),NVALS1(I),
331 & CUTOFF,NOBS2(I),NVALS2(I),ALPHA(I),DF(I)
332 6003 FORMAT (1X,A8,3X,A8,I6,I8,F13.2,I10,I8,F12.3,G15.4 /)
333 10 CONTINUE
334 WRITE (IPRINT,6004)
335 6004 FORMAT (/,' NOBS IS THE NUMBER OF NON-MISSING OBSERVATIONS ',
336 & ' IN THE ORIGINAL DATA. '//
337 & ' NVALS IS THE NUMBER OF NON-MISSING SEASONAL VALUES',
338 & ' CONSTRUCTED. ')
339 RETURN
340 END
341
342 C CUMULATIVE DISTRIBUTION FUNCTION FOR THE NORMAL DISTRIBUTION.
343 C PRIMARY REFERENCE IS ABRAMOWITZ AND STEGUN,
344 C NBS HANDBOOK OF MATHEMATICAL FUNCTIONS, EQ. 26.2.19.
345 FUNCTION CDFN(X)
346 IF (X) 10,20,30
347 10 CONTINUE
348 IF (X.LT.-6.0) GO TO 40
349 T=-X
350 CDFN=J.5/(1.0+0.0498673470*T+0.0211410061*T**2+0.0032776263*T**3
& +0.380036E-4*T**4+0.488906E-4*T**5+0.53830E-5*T**6)**16

```

```

351 RETURN
352 20 CONTINUE
353   CDFN=0.500
354 RETURN
355 30 CONTINUE
356   IF (X.GT.6.0) GO TO 50
357   CDFN=1.0-0.5/(1.0+0.0598673470*X+0.0211410061*X**2+0.0032776263*X
358   & **3+0.380036E-4*X**4+0.488906E-4*X**5+0.53830E-5*X**6)**16
359 RETURN
360 40 CONTINUE
361   CDFN=0.000
362 RETURN
363 50 CONTINUE
364   CDFN=1.000
365 RETURN
366 END
367 C CONVERT A SAS MISSING VALUE TO ONE RECOGNIZED BY THESE ROUTINES.
368 FUNCTION FIXMIS(X)
369 REAL*8 X,VALUE
370 EQUIVALENCE(VALUE,MISX)
371 VALUE=X
372 IF (VALUE.EQ.0.000.AND.MISX.NE.0) VALUE=-1.0D31
373 FIXMIS=VALUE
374 RETURN
375 END
376 SUBROUTINE RANK(X,R,N)
377 C COMPUTES THE RANKS OF THE VALUES IN VECTOR X
378 C X IS NOT RE-ARRANGED ON RETURN
379 C IN CASES OF TIES, MID-RANKS ARE USED
380 C MISSING VALUES ARE NOT ALLOWED
381 REAL X(N), R(N), Y(1200)
382 INTEGER ORD(1200)
383 C INITIALIZE ORD AND Y
384   DO 10 I = 1, N
385     ORD(I) = I
386     Y(I) = X(I)
387 10 CONTINUE
388 C REARRANGE Y IN ASCENDING ORDER
389 CALL VSRTR(Y,N,ORD)
390 C INITIAL RANKING
391   DO 100 I = 1,N
392     R(ORD(I)) = I
393 100 CONTINUE
394 C ADJUST RANKS FOR TIES
395   I = 1
396 C VALUE ALWAYS TIED WITH ITSELF
397   IEND = I
398 150 IEND = IEND + 1
399 IF (IEND.LE.N.AND.Y(IEND).EQ.Y(I)) GO TO 150
400 C COMPUTE AVERAGE RANK

```

```

401 AVE = (IEND*(IEND-1)-I*(I-1))/(2.0*(IEND-I))
402 IENDM1 = IEND - 1
403 DO 190 J = I, IENDM1
404 R(ORD(J)) = AVE
405 190 CONTINUE
406 I = IEND
407 IF(I.LT.N) GO TO 150
408 RETURN
409 END
410 //LKED.SYSIN DD *
411 ENTRY ENTRY THIS STATEMENT MUST BE PRESENT
412 SETSSI AD0J0000
413 NAME SEASRS2(R)
414 /*
415 //

```


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