

DP-7

POLYNOMIAL REGRESSION

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ABSTRACT

For any non linear function $Y = f(X)$ regression may be obtained by fitting a polynomial. The general form of the polynomial regression is as given under:

$$Y = a_0 + a_1X + a_2X^2 + \dots \dots \dots a_mX^m + \epsilon$$

where:

Y is the dependent variable and $a_0, a_1 \dots \dots a_m$ are the regression coefficients. The documentation of the computer programme for polynomial regression includes the listing of the source file, data file and output file with test data and example calculations. The details of various statistics given in the programme output have also been given in the documentation.

1.0 INTRODUCTION

Regression represents a mathematical equation expressing one random variable as being correlatively related to another random variable or several random variables. The regression equation may be any function that can be fitted to a set of observed variables. If the variables are linearly related then the regression is called linear regression. In non-linear regression the variables are non-linearly related.

For any non-linear function, $Y = f(X)$ regression may be obtained by fitting a polynomial. The general form of the polynomial regression is as given under:

$$Y = a_0 + a_1X + a_2X^2 \dots\dots\dots + a_mX^m + \epsilon \quad \dots(1)$$

The coefficients $a_0, a_1, \dots\dots\dots, a_m$ are the regression coefficients and are determined by the least squares method of parameter estimation. ϵ is the error term. The power order M is chosen so as to minimise the sum of the squares of deviations from the line. The power M should be much lower than the sample size, N , in order to have a sufficient number of degrees of freedom ($N - M - 1$) and to have a reliable estimate of the standard deviation. Generally, the value of M is between 2 and 4 as it is very difficult to explain higher degree of M .

It is preferable to plot the data on a single graph first to have a preliminary idea about the value of M or the

degree of polynomial to be fitted to the data and also to eliminate any inconsistent points.

The polynomial regression analysis is generally used for trend analysis in hydrologic data. The programme for polynomial regression analysis described in this documentation has been taken from IBM Scientific Subroutine Package and implemented/tested on VAX-11/780 computer system of National Institute of Hydrology, Roorkee.

2.0 PURPOSE OF THE PROGRAMME

The programme calls subroutines to perform the regression analysis. The programme prints the regression coefficients and analysis of variance tables for polynomials of successively increasing degrees. The programme also optionally prints the table of residuals and a plot of observed Y values and Y estimates versus base variable X. If there is no reduction in the residual sum of squares between two successive degrees of polynomials the programme terminates the problem, otherwise it continues till the analysis for the highest degree of polynomial specified is completed.

3.0 SPECIFIC METHOD

Various statistical parameters given in the output are computed by the following equations:

- a) Regression coefficients for successive degrees of polynomials:

Regression coefficients are the values of a_1 , a_2 , a_3 etc. in the equation given below:

$$Y = a_0 + a_1X + a_2X^2 + \dots + a_mX^m \quad \dots (2)$$

For first degree polynomial there will be only one regression coefficient. The value of a_0 is the intercept.

- b) Analysis of variance for successive degrees of polynomials:

Analysis of variance includes computation of variance explained due to regression and due to deviation about the regression line. The F-value is also computed.

- i) Variance due to regression is given by the following expression:

$$\text{Variance due to regression} = \sum_{i=1}^N (Y'_i - \bar{Y})^2 \quad \dots (3)$$

where:

- Y'_i : Estimated i^{th} value of Y
 \bar{Y} : The mean of Y values

- ii) Variance due to deviation about regression is given by:

$$\text{Variance due to deviation about regression} = \sum_{i=1}^N (Y_i - Y'_i)^2 \dots (4)$$

where:

Y_i : Observed i^{th} value of Y

Y'_i : Estimated i^{th} value of Y

c) F value :

F value is the ratio of mean squares due to regression and mean squares about the regression line.

$$F \text{ value} = \frac{\text{Mean squares due to regression}}{\text{Mean squares about regression}} \dots (5)$$

4.0 COMPUTER PROGRAMME

The main features of computer programme and different subroutines are discussed as follows:

4.1 Programme Subroutines

The computer programme for polynomial regression consists of the main routine named PREG and five other subroutines named GDATA for data matrix generation for polynomial regression, ORDER for rearrangement of inter-correlations, MINV for matrix inversion, MULTR for multiple linear regression and PLOT for plotting. These subroutines are described in following paragraphs:

A. Subroutine GDATA (N, M, X, XBAR, STD, D, SUMSQ)

This subroutine generates independent variables upto the M^{th} power (the highest degree polynomial specified) and calculates means, standard deviations, sums of cross products of deviations from means and product moment correlation coefficients. The calling arguments are:

N : Number of observations

M : The highest degree polynomial to be fitted

X : Input matrix ($N \times M+1$). When the subroutine is called, data for the independent variable are stored in the first column of the matrix, and data for the dependent variable are stored in the last column of the matrix. Upon returning

to the calling routine, generated powers of the independent variable are stored in the columns 2 through M

XBAR : Output vector of length $M + 1$ containing means of independent and dependent variables

STD: Output vector of length $M+1$ containing standard deviations of independent and dependent variables

D : Output matrix containing correlation coefficients

SUMSQ: Output vector of length $(M+1)$ containing sums of products of deviations from means of independent and dependent variable

B. Subroutine ORDER (M,R, NDEP, K, ISAVE, RX, RY)

The purpose of this subroutine is to construct from larger matrix of correlation coefficients, a subset of matrix of intercorrelations among independent variables and a vector of intercorrelations of independent variables with dependent variable. The calling arguments are:

M : Number of variables and order of matrix R

R : Input matrix containing correlation coefficients

NDEP: The subscript number of the dependent variable

K : Number of independent variables to be included in the forthcoming regression, K must be greater than or equal to 1.

ISAVE: Input vector of length $K + 1$ containing, in ascending order, the subscript numbers of K independent variables to be included in the forthcoming regression upon returning to the calling routine. This vector contains, in addition, the subscript number of the dependent variable in $K+1$ position

RX: Output matrix ($K \times K$) containing inter-correlations among independent variables to be used in forthcoming regression

RY: Output vector of length K containing inter-correlations of independent variables with dependent variable

C. Subroutine MINV (A, N, D, L, M)

The subroutine is used for matrix inversion. The calling arguments are:

A : Input matrix destroyed in computation and replaced by resultant inverse

N : Order of matrix A

D : Resultant determinant

L : Work vector of length N

M : Work vector of length N

D. Subroutine MULTR (N, K, XBAR, STD, D, RX, RY, ISAVE, B, SB, T, ANS)

This performs a multiple linear regression analysis for a dependent variable and a set of independent variables.

The calling arguments are :

N : Number of observations

K : Number of independent variables in the regression

XBAR : Input vector of length M containing means of all variables. M is the number of variables in the observations

STD: Input vector of length M containing standard deviations of all variables

D : Input vector of length M containing the diagonal of the matrix of sums of cross products of deviations from means for all variables

RX: Input matrix (K x K) containing the inverse of intercorrelations among independent variables

RY: Input vector of length K containing intercorrelations of independent variables with dependent variable

ISAVE: Input vector of length (K+1) containing subscripts of dependent variables in ascending order. The subscript of the dependent variable is stored in the last K+1 position

B : Output vector of length K containing regression coefficients

SB : Output vector of length K containing standard deviations of regression coefficients

T : Output vector of length K containing t values

ANS: Output vector of length 10 containing the following information

ANS(1) : Intercept

ANS(2) : Multiple correlation coefficient
ANS(3) : Standard error of estimate
ANS(4) : Sum of squares attributable to regression
(SSAR)
ANS(5) : Degree of freedom associated with SSAR
ANS(6) : Mean squares of SSAR
ANS(7) : Sum of squares of deviations from regression
(SSDR)
ANS(8) : Degrees of freedom associated with (SSDR)
ANS(9) : Mean squares of SSDR
ANS(10) : F value

E. Subroutine PLOT (NO, A, N, M, NL, NS)

The purpose of this subroutine is to plot Y values and Y estimates versus base variable X. The calling arguments are :

NO : Chart number
A : Matrix of data to be plotted. First column represents base variable and successive columns are the cross variables (maximum 9)
N : Number of rows in A
M : Number of columns in matrix A (equal to the total number of variables) maximum is 10
NL : Number of lines in the plot. If 0 is specified 50 lines are used
NS : Code for sorting the base variable data in ascending order
0 : Sorting is not necessary

1 = Sorting is necessary

The listing of the source programme has been given in Appendix I.

4.2 Programme Modifications

The programme capacity can be increased or decreased by making changes in dimension statements. The following are the general rules for the programme modifications:

- a. The dimension of array X must be greater than or equal to the product $N (M + 1)$, where N is the number of observations and M is the highest degree polynomial to be fitted.
- b. The dimensions of array DI must be greater than or equal to the product of $M \times M$.
- c. The dimension of array D must be greater than or equal to $(M+2) (M+1)/2$.
- d. The dimensions of arrays B, E, SB and T must be greater than or equal to the highest degree polynomial to be fitted, M.
- e. The dimension of arrays XBAR, STD, COE, SUMSQ and ISAVE must be greater than or equal to $(M+1)$.
- f. The dimension of array P must be greater than or equal to $3 \times N$.

5.0 INPUT SPECIFICATIONS, OUTPUT DESCRIPTION AND
RESTRICITIONSON USE

5.1 Input Specifications

Input data file contains control and data lines/cards.

5.1.1 Control cards

In first line, title of the problem is given in A
format.

Second line

Column

Contents

1-5

Number of observations

6-7

Highest degree polynomial to be
fitted

8

Option code for plotting Y
values and Y estimates versus
base variable X

0 if it is not required

1 if it is required

5.1.2 Data cards

Input data are read into the computer one observation
at a time i.e. each pair of X and Y values in free format.

5.2 Output Description

The output for the programme for polynomial regression
includes:

- a. Regression coefficients for successive degree polynomials

- b. Analysis of variance for successive degree polynomials
- c. Table of residuals for the final degree polynomial
- d. Plot of Y values and Y estimates versus base variable X (Optional). Observed Y values are denoted by 1 and estimated values by 2.

5.3 Restrictions on Use

All the pairs should be complete i.e. there should not be any missing data.

6.0 TEST DATA

The programme for polynomial regression has been run on 93 years (1887-1979) annual peak stage data of river Narmada at BROACH. This has been done in order to see the trend in the data. The stage data is given below:

Sl No.	Year	Stage (m)	Sl No.	Year	Stage (m)	Sl No.	Year	Stage (m)
1.	1887	26.50	35.	1921	24.50	69.	1955	28.50
2.	1888	21.00	36.	1922	24.50	70.	1956	27.00
3.	1889	23.50	37.	1923	30.00	71.	1957	27.50
4.	1890	25.00	38.	1924	27.00	72.	1958	25.50
5.	1891	22.50	39.	1925	22.00	73.	1959	33.00
6.	1892	27.50	40.	1926	29.50	74.	1960	25.50
7.	1893	24.50	41.	1927	28.00	75.	1961	25.20
8.	1894	30.00	42.	1928	27.50	76.	1962	32.30
9.	1895	21.50	43.	1929	25.00	77.	1963	23.00
10.	1896	30.00	44.	1930	30.50	78.	1964	25.25
11.	1897	27.00	45.	1931	29.50	79.	1965	23.00
12.	1898	30.00	46.	1932	29.00	80.	1966	20.60
13.	1899	18.25	47.	1933	34.00	81.	1967	23.75
14.	1900	28.00	48.	1934	29.50	82.	1968	38.50
15.	1901	26.50	49.	1935	23.50	83.	1969	29.60
16.	1902	20.50	50.	1936	26.50	84.	1970	41.50
17.	1903	23.50	51.	1937	30.50	85.	1971	26.00
18.	1904	21.50	52.	1938	30.50	86.	1972	32.00
19.	1905	30.70	53.	1939	34.00	87.	1973	36.50
20.	1906	32.50	54.	1940	26.50	88.	1974	29.98
21.	1907	25.00	55.	1941	29.00	89.	1975	28.50
22.	1908	24.50	56.	1942	34.50	90.	1976	26.50
23.	1909	17.00	57.	1943	25.50	91.	1977	26.99
24.	1910	25.50	58.	1944	35.00	92.	1978	29.54
25.	1911	20.50	59.	1945	33.00	93.	1979	31.49
26.	1912	28.00	60.	1946	27.50			
27.	1913	33.00	61.	1947	31.50			
28.	1914	21.50	62.	1948	29.50			
29.	1915	24.50	63.	1949	29.00			
30.	1916	27.50	64.	1950	35.00			
31.	1917	27.00	65.	1951	23.50			
32.	1918	19.50	66.	1952	23.50			
33.	1919	32.50	67.	1953	33.00			
34.	1920	22.50	68.	1954	29.50			

7.0 EXAMPLE CALCULATIONS

The calculations for first degree polynomial are given below:

Step 1 : For first degree polynomial (straight line)

regression coefficients are obtained by method of least squares. The coefficients are given below:

$$\text{Intercept} = 24.62$$

$$\text{Regression coefficient } a_1 = 0.0611$$

$$\text{Step 2 : Sum of squares due to regression} = \sum_{i=1}^{93} (Y_i' - \bar{Y}_i)^2$$

$$= 250.37$$

Sum of squares due to deviation from regression

$$= \sum_{i=1}^{93} (Y_i' - Y_i)^2 = 1615.39$$

$$F \text{ value} = \frac{\text{Mean squares due to regression}}{\text{Mean squares due to deviation from regression}}$$

$$= \frac{(250.37/2)}{(1615.39/91)} = 14.10$$

8.0 APPLICATION, SAMPLE INPUT AND SAMPLE OUTPUT

The programme for polynomial regression has been run on the 93 years peak stage data of river Narmada at Broach to analyse the trend in the data.

8.1 Sample input

The listing of the sample input (data file) has been given in Appendix II.

In the sample problem total number of observations is 93 and the highest degree of polynomial required is 2. The plot is also required. The value of N, M and N PLOT will be 93, 2, 1 respectively.

In the data lines, the value of X will vary from 1 to 93. The X values and Y values have been punched in free format.

8.2 Sample Output

The listing of the sample output has been given in Appendix III.

The programme first fits one degree polynomial to the stages. The relation obtained is

$$Y = 24.623 + 0.0611207 t$$

The deviation about regression line is 1615.3916. In case of second degree polynomial, the relation obtained is

$$Y = 23.83937 + 0.11062 t - 0.0005267 t^2$$

The deviation about regression line is 1604.6779. The improvement in terms of sum of squares for second degree polynomial is 10.71358. So second degree polynomial better fits the data as compared to first degree. The table of residuals and plot of Y values denoted by (1) and Y estimates denoted by (2) versus base variable X has been given in Appendix III.

9.0 RECOMMENDATIONS

The programme for polynomial regression can deal upto 100 observations and 10^{th} degree polynomial. Therefore if a problem satisfies the above conditions it is not necessary to modify the programme. However, if there are more than 100 observations or if greater than 10^{th} degree polynomial is desired, dimension statements in the main programme must be modified to handle the problem.

REFERENCES

1. Haan, C.T. (1977), 'Statistical Methods in Hydrology', Iowa State University Press, Ames, Iowa.
2. Scientific Subroutine Package, International Business Machines, White Plains, N.Y.

APPENDIX I

POLYNOMIAL REGRESSION PROGRAMME

```

C   MASTER POLYNOMIAL REGRESSION
    DIMENSION X(1100),BI(100),B(66),R(10),E(10),SB(10),
    IT(10),XBAR(11),STD(11),COE(11),SUMSQ(11),ISAVE(11),
    ZANS(10),P(300),TITLE(80)
    OPEN(UNIT=5,FILE='PREG.DAT',STATUS='OLD')
    OPEN(UNIT=6,FILE='PREG.OUT',STATUS='NEW')
1   FORMAT(I5,I2,I1)
2   FORMAT(2F6.0)
3   FORMAT(1X,' POLYNOMIAL REGRESSION.....')
4   FORMAT(/,' NUMBER OF OBSERVATIONS',I6/)
5   FORMAT(/' POLYNOMIAL REGRESSION OF DEGREE',I3)
6   FORMAT(' INTERCEPT',F20.4)
7   FORMAT(/,' REGRESSION COEFFICIENTS'/(6F20.4))
8   FORMAT(1H0/20X,' ANALYSIS OF VARIANCE FOR',I4,
    1'DEGREE OF POLYNOMIAL')
9   FORMAT(1H0,1X,' SOURCE OF VARIATION',5X,' DEGREE OF'
    1,3X,' SUM OF',3X,4HMEAN,5X,1HF,4X,' IMPROVEMENT IN'
    2TERMS'/26X,' FREEDOM',4X,' SQUARES',3X,' SQUARE',3X,
    3SHVALUE,1X,' OF SUM OF SQUARES')
10  FORMAT(' DUE TO REGRESSION',7X,I6,F13.2
    1,F10.2,F7.2,F12.2)
11  FORMAT(' DEVIATION ABOUT REGRESSION',I5,F13.2,F10.2)
12  FORMAT(5X,' TOTAL',11X,I6,F13.2//)
13  FORMAT(' NO IMPROVEMENT')
14  FORMAT(1H0//27X,' TABLE OF RESIDUALS')
    116 HOBSERVATION NO.,5X,7HX VALUE,7X,7HY VALUE,7X,
    210HY ESTIMATE,7X,8HRESIDUAL/)
15  FORMAT(3X,I6,F18.5,F14.5,F17.5,F15.5)
16  FORMAT(80A1)
100 READ(5,16)TITLE
    READ(5,1)N,M,NPLOT
    WRITE(6,16)TITLE
    WRITE(6,3)
    WRITE(6,4)N
C   READ INPUT DATA
    L=N*M
    DO 110 I=1,N
        J=L+I
110  READ(5,*) X(I),X(J)
        CALL GDATA(N,M,X,XBAR,STD,B,SUMSQ)
        MM=M+1
        SUM=0.0
        NT=N-1
        DO 200 I=1,M
            ISAVE(I)=I
            CALL ORDER(MM,B,MM,I,ISAVE,RI,E)
            CALL MINV(RI,I,BET,B,T)
            CALL MULTR(N,I,XBAR,STD,SUMSQ,B,E,ISAVE,B,SB,I,ANS)
            WRITE(6,5) I
            IF(ANS(7)) 140,130,130

```

```

130  SUMIP=ANS(4)-SUM
    IF(SUMIP) 140,140,150
140  WRITE(6,13)
    GO TO 210
150  WRITE(6,6)ANS(1)
    WRITE(6,7)(B(J),J=1,I)
    WRITE(6,8) I
    WRITE(6,9)
    SUM=ANS(4)
    WRITE(6,10) I,ANS(4),ANS(6),ANS(10),SUMIP
    NI=ANS(8)
    WRITE(6,11) NI,ANS(7),ANS(9)
    WRITE(6,12) NT,SUMSR(M)
    COE(1)=ANS(1)
    DO 160 J=1,I
160  COE(J+1)=B(J)
    LA=I
200  CONTINUE
C    TEST WHETHER PLOT IS REQUIRED OR NOT
210  IF(NPLOT) 100,100,220
220  NP3=N+N
    DO 230 I=1,N
    NP3=NP3+1
    P(NP3)=COE(1)
    L=I
    DO 230 J=1,LA
    P(NP3)=P(NP3)+X(L)*COE(J+1)
230  L=L+N
    N2=N
    L=N*M
    DO 240 I=1,N
    P(I)=X(I)
    N2=N2+1
    L=L+1
240  P(N2)=X(L)
    WRITE(6,3)
    WRITE(6,5) LA
    WRITE(6,14)
    NP2=N
    NP3=N+N
    DO 250 I=1,N
    NP2=NP2+1
    NP3=NP3+1
    RESID=P(NP2)-P(NP3)
250  WRITE(6,15) I,P(I),P(NP2),P(NP3),RESID
    CALL PLOT(LA,P,N,3,0,0)
    STOP
    END
    SUBROUTINE ORDER(M,R,NDEF,K,ISAVE,RX,RY)
    DIMENSION R(1),ISAVE(1),RX(1),RY(1)

```



```

MM=0
DO 130 J=1,K
L2=ISAVE(J)
IF(NDEF-L2)122,123,123
122 L=NDEF+(L2*L2-L2)/2
GO TO 125
123 L=L2+(NDEF*NDEF-NDEF)/2
125 RY(J)=R(L)
DO 130 I=1,K
L1=ISAVE(I)
IF(L1-L2)127,128,128
127 L=L1+(L2*L2-L2)/2
GO TO 129
128 L=L2+(L1*L1-L1)/2
129 MM=MM+1
130 RX(MM)=R(L)
ISAVE(K+1)=NDEF
RETURN
END

```

C

```

SUBROUTINE MINV(A,N,D,L,M)
DIMENSION A(1),L(1),M(1)

```

C

```

SEARCH FOR LARGEST ELEMENT

```

```

D=1.0
NK=-N
DO 80 K=1,N
NK=NK+N
L(K)=K
M(K)=K
KK=NK+K
BIGA=A(KK)
DO 20 J=K,N
IZ=N*(J-1)
DO 20 I=K,N
IJ=IZ+I

```

10

```

IF(ABS(BIGA)-ABS(A(IJ))) 15,20,20

```

15

```

BIGA=A(IJ)

```

```

L(K)=I

```

```

M(K)=J

```

20

```

CONTINUE

```

C

```

INTERCHANGE ROWS

```

```

J=L(K)

```

```

IF(J-K) 35,35,25

```

25

```

KI=K-N

```

```

DO 30 I=1,N

```

```

KI=KI+N

```

```

HOLD=-A(KI)

```

```

JI=KI-K+J

```

```

A(KI)=A(JI)

```

30

```

A(JI)=HOLD

```

```

C      INTERCHANGE COLUMNS
35     I=M(K)
      IF(I-K)45,45,38
38     JP=N*(I-1)
      DO 40 J=1,N
      JK=NK+J
      JI=JP+J
      HOLD=-A(JK)
      A(JK)=A(JI)
40     A(JI)=HOLD
C      DIVIDE COLUMNS BY MINUS PIVOT
45     IF(BIGA)48,46,48
46     D=0.0
      RETURN
48     DO 55 I=1,N
      IF(I-K)50,55,50
50     IK=NK+I
      A(IK)=A(IK)/(-BIGA)
55     CONTINUE
C      REDUCE MATRIX
      DO 65 I=1,N
      IK=NK+I
      HOLD=A(IK)
      IJ=I-N
      DO 65 J=1,N
      IJ=IJ+N
      IF(I-K)60,65,60
60     IF(J-K)62,65,62
62     KJ=IJ-I+K
      A(IJ)=HOLD*A(KJ)+A(IJ)
65     CONTINUE
C      DIVIDE ROW BY PIVOT
      KJ=K-N
      DO 75 J=1,N
      KJ=KJ+N
      IF(J-K)70,75,70
70     A(KJ)=A(KJ)/BIGA
75     CONTINUE
C      PRODUCT OF PIVOTS
      D=D*BIGA
C      REPLACE PIVOT BY RECIPROCAL
      A(KK)=1.0/BIGA
80     CONTINUE
C      FINAL ROW AND COLUMN INTERCHANGE
      K=N
100    K=(K-1)
      IF(K)150,150,105
105    I=L(K)
      IF(I-K)120,120,108
108    JQ=N*(K-1)

```

```

JR=N*(I-1)
DO 110 J=1,N
JK=J0+J
HOLD=A(JK)
JI=JR+J
A(JK)=-A(JI)
110 A(JI)=HOLD
120 J=M(K)
IF(J-K)100,100,125
125 KI=K-N
DO 130 I=1,N
KI=KI+N
HOLD=A(KI)
JI=KI-K+J
A(KI)=-A(JI)
130 A(JI)=HOLD
GO TO 100
150 RETURN
END
C
SUBROUTINE MULTR(N,K,XBAR,STD,D,RX,NY,ISAVE,B,SB,T,ANS)
DIMENSION XBAR(1),STD(1),RX(1),RY(1),D(1),ISAVE(1),
1B(1),SB(1),T(1),ANS(1)
MM=K+1
C
BETA WEIGHTS
DO 100 J=1,K
100 B(J)=0.0
DO 110 J=1,K
L1=K*(J-1)
DO 110 I=1,K
L=L1+I
110 B(J)=B(J)+RY(I)*RX(L)
RM=0.0
RO=0.0
L1=ISAVE(MM)
C
COEFFICIENTS OF DETERMINATION
DO 120 I=1,K
RM=RM+B(I)*RY(I)
C
REGRESSION COEFFICIENTS
L=ISAVE(I)
B(I)=B(I)*(STD(L1)/STD(L))
C
INTERCEPT
120 RO=RO+B(I)*XBAR(L)
RO=XBAR(L1)-RO
C
SUM OF SQUARES ATTRIBUTABLE TO REGRESSION
SSAR=RM*D(L1)
C
MULTIPLE CORRELATION COEFFICIENT
122 RM=SQRT(ABS(RM))
C
SUM OF SQUARES OF DEVIATIONS FROM REGRESSION
SSDR=D(L1)-SSAR

```

```

C      VARIANCE OF ESTIMATE
      FN=N-K-1
      SY=SSDR/FN
C      STANDARD DEVIATIONS OF REGRESSION COEFFICIENTS
      DO 130 J=1,K
      L1=K*(J-1)+J
      L=ISAVE(J)
125    SB(J)=SQRT(ABS((RX(L1)/D(L))*SY))
C      COMPUTED T-VALUES
130    T(J)=B(J)/SB(J)
C      STANDARD ERROR OF ESTIMATE
135    SY=SQRT(ABS(SY))
C      F VALUE
      FK=K
      SSARM=SSAR/FK
      SSDRM=SSDR/FN
      F=SSARM/SSDRM
      ANS(1)=B0
      ANS(2)=RM
      ANS(3)=SY
      ANS(4)=SSAR
      ANS(5)=FK
      ANS(6)=SSARM
      ANS(7)=SSDR
      ANS(8)=FN
      ANS(9)=SSDRM
      ANS(10)=F
      RETURN
      END
      SUBROUTINE PLOT(ND,A,N,M,NL,NS)
      DIMENSION OUT(101),YPR(11),ANG(9),A(300)
1      FORMAT(1H1,60X,7H CHART ,I3//)
2      FORMAT(1H ,F11.4,5X,101A1)
3      FORMAT(1H )
4      FORMAT(10H 123456789)
5      FORMAT(10A1)
7      FORMAT(1H ,16X,101H .
          1      .
          1      .)
8      FORMAT(1H0,9X,11F10.4)
      NLL=NL
      IF(NS)16,16,10
C      SORT BASE VARIABLE IN ASCENDING ORDER
10     DO 15 I=1,N
        DO 14 J=1,N
          IF(A(I)-A(J)) 14,14,11
11     L=I-N
        LL=J-N
        DO 12 K=1,M
          L=L+N

```

```

        LL=LL+N
        F=A(L)
        A(L)=A(LL)
12      A(LL)=F
14      CONTINUE
15      CONTINUE
C TEST NLL
16      IF(NLL) 20,18,20
18      NLL=50
20      WRITE(6,1) NO
        REWIND 13
        WRITE(13,4)
        REWIND 13
        READ(13,5) BLANK,(ANG(I),I=1,9)
        REWIND 13
C FIND SCALE FOR BASE VARIABLE
        XSCAL=(A(N)-A(1))/(FLOAT(NLL-1))
C FIND SCALE FOR CROSS VARIABLE
        M1=N+1
        YMIN=A(M1)
        YMAX=YMIN
        M2=M*N
        DO 40 J=M1,M2
        IF(A(J)-YMIN) 28,26,26
26      IF(A(J)-YMAX) 40,40,30
28      YMIN=A(J)
        GOTO 40
30      YMAX=A(J)
40      CONTINUE
        YSCAL=(YMAX-YMIN)/100.0
C FIND BASE VARIABLE PRINT POSITION
        XB=A(1)
        L=1
        MY=M-1
        I=1
45      F=I-1
        XPR=XB+F*XSCAL
        IF(A(L)-XPR) 50,50,70
50      DO 55 IX=1,101
55      OUT(IX)=BLANK
        DO 60 J=1,MY
        LL=L+J*N
        JP=((A(LL)-YMIN)/YSCAL)+1.0
        OUT(JP)=ANG(J)
60      CONTINUE
        WRITE(6,2)XPR,(OUT(IZ),IZ=1,101)
        L=L+1
        GOTO 80
70      WRITE(6,3)
80      I=I+1

```

```

      IF(I-NLL)45,84,86
84   XPR=A(N)
      GOTO 50
86   WRITE(6,7)
      YPR(1)=YMIN
      DO 90 KN=1,9
90   YPR(KN+1)=YPR(KN)+YSCAL*10.0
      YPR(11)=YMAX
      WRITE(6,8)(YPR(IP),IP=1,11)
      RETURN
      END
      SUBROUTINE GDATA(N,M,X,XBAR,STD,D,SUMSQ)
      DIMENSION X(1),XBAR(1),STD(1),D(1),SUMSQ(1)
      IF(M-1) 105,105,90
90   L1=0
      DO 100 I=2,M
      L1=L1+N
      DO 100 J=1,N
      L=L1+J
      K=L-N
100  X(L)=X(K)*X(J)
105  MM=M+1
      DF=N
      L=0
      DO 115 I=1,MM
      XBAR(I)=0.
      DO 110 J=1,N
      L=L+1
110  XBAR(I)=XBAR(I)+X(L)
115  XBAR(I)=XBAR(I)/DF
      DO 130 I=1,MM
130  STD(I)=0.
      L=((MM+1)*MM)/2
      DO 150 I=1,L
150  B(I)=0.
      DO 170 K=1,N
      L=0
      DO 170 J=1,MM
      L2=N*(J-1)+K
      T2=X(L2)-XBAR(J)
      STD(J)=STD(J)+T2
      DO 170 I=1,J
      L1=N*(I-1)+K
      T1=X(L1)-XBAR(I)
      L=L+1
170  D(L)=D(L)+T1*T2
      L=0
      DO 175 J=1,MM
      DO 175 I=1,J
      L=L+1

```

```

175  D(L)=D(L)-STD(I)*STD(J)/DF
      L=L+1
      DO 180 I=1,MM
        L=L+I
        SUMSQ(I)=D(L)
180  STD(I)=SQRT(ABS(D(L)))
C    CALCULATE CORRELATION COEFFICIENT
      L=L+1
      DO 190 J=1,MM
        DO 190 I=1,J
          L=L+1
190  D(L)=D(L)/(STD(I)*STD(J))
C    CALCULATE STANDARD DEVIATION
      DF=SQRT(DF-1.0)
      DO 200 I=1,MM
200  STD(I)=STD(I)/DF
      RETURN
      END

```

APPENDIX II

TEST INPUT

93 YEARS STAGE DATA AT BROACH
00093021

1.	26.50
2.	21.00
3.	23.50
4.	25.00
5.	22.50
6.	27.50
7.	24.50
8.	30.00
9.	21.50
10.	30.00
11.	27.00
12.	30.00
13.	18.25
14.	28.00
15.	26.50
16.	20.50
17.	23.50
18.	21.50
19.	30.70
20.	32.50
21.	25.00
22.	24.50
23.	17.00
24.	25.50
25.	20.50
26.	28.00
27.	33.00
28.	21.50
29.	24.50
30.	27.50
31.	27.00
32.	19.50
33.	32.50
34.	22.50
35.	24.50
36.	24.50
37.	30.00
38.	27.00
39.	22.00
40.	29.50
41.	28.00
42.	27.50
43.	25.00
44.	30.50
45.	29.50
46.	29.00
47.	34.00
48.	29.50

49.	23.50
50.	26.50
51.	30.50
52.	30.50
53.	34.00
54.	26.50
55.	29.00
56.	34.50
57.	25.50
58.	35.00
59.	33.00
60.	27.50
61.	31.50
62.	29.50
63.	29.00
64.	35.00
65.	23.50
66.	23.50
67.	33.00
68.	29.50
69.	28.50
70.	27.00
71.	27.50
72.	25.50
73.	33.00
74.	25.50
75.	25.20
76.	32.30
77.	23.00
78.	25.25
79.	23.00
80.	20.60
81.	23.75
82.	38.50
83.	29.60
84.	41.50
85.	26.00
86.	32.00
87.	36.50
88.	29.98
89.	28.50
90.	26.50
91.	26.99
92.	29.54
93.	31.49

APPENDIX III

TEST OUTPUT

93 YEARS STAGE DATA AT BROACH
POLYNOMIAL REGRESSION.....

NUMBER OF OBSERVATIONS 93

POLYNOMIAL REGRESSION OF DEGREE 1

INTERCEPT 24.6236

REGRESSION COEFFICIENTS

0.0611

0

SOURCE OF VARIATION	ANALYSIS OF VARIANCE FOR		1DEGREE OF POLYNOMIAL		
	DEGREE OF FREEDOM	SUM OF SQUARES	MEAN SQUARE	F VALUE	IMPROVEMENT INTERMS VALUE OF SUM OF SQUARES
DUE TO REGRESSION	1	250.38	250.38	14.10	250.38
DEVIATION ABOUT REGRESSION	91	1615.39	17.75		
TOTAL	92	1865.77			

POLYNOMIAL REGRESSION OF DEGREE 2

INTERCEPT 23.8397

REGRESSION COEFFICIENTS

0.1106

-0.0005

0

SOURCE OF VARIATION	ANALYSIS OF VARIANCE FOR		2DEGREE OF POLYNOMIAL		
	DEGREE OF FREEDOM	SUM OF SQUARES	MEAN SQUARE	F VALUE	IMPROVEMENT INTERMS VALUE OF SUM OF SQUARES
DUE TO REGRESSION	2	261.09	130.55	7.32	10.71
DEVIATION ABOUT REGRESSION	90	1604.68	17.83		
TOTAL	92	1865.77			

POLYNOMIAL REGRESSION.....

POLYNOMIAL REGRESSION OF DEGREE 2

0

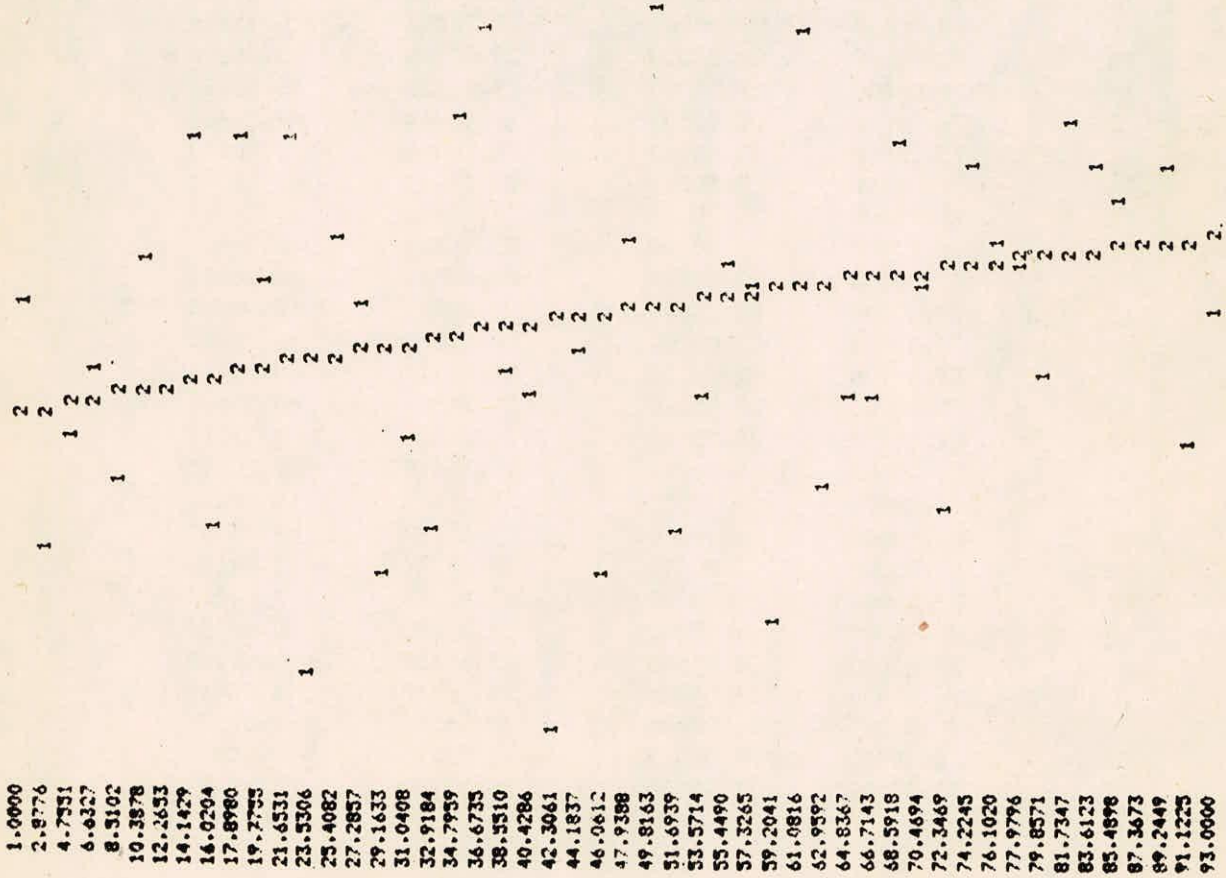
TABLE OF RESIDUALS

OBSERVATION NO.:	X VALUE	Y VALUE	Y ESTIMATE	RESIDUAL
1	1.00000	26.50000	23.94983	2.55017
2	2.00000	21.00000	24.05888	-3.05888

3	3.00000	23.50000	24.16687	-0.66687
4	4.00000	25.00000	24.27381	0.72619
5	5.00000	22.50000	24.37969	-1.87969
6	6.00000	27.50000	24.48453	3.01547
7	7.00000	24.50000	24.58831	-0.08831
8	8.00000	30.00000	24.69103	5.30897
9	9.00000	21.50000	24.79271	-3.29271
10	10.00000	30.00000	24.89333	5.10667
11	11.00000	27.00000	24.99289	2.00711
12	12.00000	30.00000	25.09141	4.90859
13	13.00000	18.25000	25.18887	-6.93887
14	14.00000	28.00000	25.28527	2.71473
15	15.00000	26.50000	25.38063	1.11937
16	16.00000	20.50000	25.47493	-4.97493
17	17.00000	23.50000	25.56817	-2.06817
18	18.00000	21.50000	25.66037	-4.16037
19	19.00000	30.70000	25.75151	4.94849
20	20.00000	32.50000	25.84159	6.65841
21	21.00000	25.00000	25.93063	-0.93063
22	22.00000	24.50000	26.01861	-1.51861
23	23.00000	17.00000	26.10553	-9.10553
24	24.00000	25.50000	26.19141	-0.69141
25	25.00000	20.50000	26.27623	-5.77623
26	26.00000	28.00000	26.35999	1.68001
27	27.00000	33.00000	26.44271	6.55729
28	28.00000	21.50000	26.52437	-5.02437
29	29.00000	24.50000	26.60498	-2.10498
30	30.00000	27.50000	26.68453	0.81547
31	31.00000	27.00000	26.76303	0.26697
32	32.00000	19.50000	26.84048	-7.34048
33	33.00000	32.50000	26.91687	5.58313
34	34.00000	22.50000	26.99221	-4.49221
35	35.00000	24.50000	27.06650	-2.56650
36	36.00000	24.50000	27.13973	-2.63973
37	37.00000	30.00000	27.21191	2.78809
38	38.00000	27.00000	27.28304	-0.28304
39	39.00000	22.00000	27.35311	-5.35311
40	40.00000	29.50000	27.42213	2.07787
41	41.00000	28.00000	27.49010	0.50990
42	42.00000	27.50000	27.55701	-0.05701
43	43.00000	25.00000	27.62288	-2.62288
44	44.00000	30.50000	27.68768	2.81232
45	45.00000	29.50000	27.75144	1.74856
46	46.00000	29.00000	27.81414	1.18586
47	47.00000	34.00000	27.87578	6.12422
48	48.00000	29.50000	27.93638	1.56362
49	49.00000	23.50000	27.99592	-4.49592
50	50.00000	26.50000	28.05441	-1.55441
51	51.00000	30.50000	28.11184	2.38816
52	52.00000	30.50000	28.16822	2.33178

53	53.00000	34.00000	28.22355	5.77645
54	54.00000	26.50000	28.27782	-1.77782
55	55.00000	29.00000	28.33104	0.66896
56	56.00000	34.50000	28.38321	6.11679
57	57.00000	25.50000	28.43432	-2.96462
58	58.00000	35.00000	28.48438	6.51562
59	59.00000	33.00000	28.53339	4.46661
60	60.00000	27.50000	28.58135	-1.08135
61	61.00000	31.50000	28.62825	2.8/175
62	62.00000	29.50000	28.67409	0.82591
63	63.00000	29.00000	28.71889	0.28111
64	64.00000	35.00000	28.76263	6.23737
65	65.00000	23.50000	28.80532	-5.30532
66	66.00000	23.50000	28.84695	-5.34695
67	67.00000	33.00000	28.88753	4.11247
68	68.00000	29.50000	28.92706	0.57294
69	69.00000	28.50000	28.96553	-0.46553
70	70.00000	27.00000	29.00295	-2.00295
71	71.00000	27.50000	29.03932	-1.53932
72	72.00000	25.50000	29.07464	-3.57464
73	73.00000	33.00000	29.10890	3.89110
74	74.00000	25.50000	29.14211	-3.64211
75	75.00000	25.20000	29.17426	-3.97426
76	76.00000	32.30000	29.20536	3.09464
77	77.00000	23.00000	29.23541	-6.23541
78	78.00000	25.25000	29.26440	-4.01440
79	79.00000	23.00000	29.29235	-6.29235
80	80.00000	20.60000	29.31923	-8.71923
81	81.00000	23.75000	29.34507	-5.59507
82	82.00000	38.50000	29.36985	9.13015
83	83.00000	29.60000	29.39358	0.20642
84	84.00000	41.50000	29.41625	12.08375
85	85.00000	26.00000	29.43787	-3.43787
86	86.00000	32.00000	29.45844	2.54156
87	87.00000	36.50000	29.47796	7.02204
88	88.00000	29.98000	29.49642	0.48358
89	89.00000	28.50000	29.51383	-1.01383
90	90.00000	26.50000	29.53018	-3.03018
91	91.00000	26.99000	29.54548	-2.55548
92	92.00000	29.54000	29.55973	-0.01973
93	93.00000	31.49000	29.57292	1.91708

CHART 2



17.0000 19.4500 21.9000 24.3500 26.8000 29.2500 31.7000 34.1500 36.6000 39.0500 41.5000