

APPLIED REGRESSION IN THE PRESENCE OF X ERROR

Fourth Interim Report

Project 7788

Development of Statistical Procedures

Prepared By

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16. Abstract Regression analysis is frequently used in the engineering field to develop mathematical models for a wide variety of applications. Of the several assumptions upon which regression is based, one of the most fundamental is that the X values are known exactly and that any error is associated only with the Y measurements. Since this is not the case for many engineering applications, a study was conducted to (a) determine the magnitude of this problem and to (b) develop and test a software package that incorporates a theoretical solution found in the literature. Computer simulation is used to demonstrate both the seriousness of the problem and the effectiveness of the solution. An example based on early-strength tests of concrete demonstrates the use of the new software. The complete Fortran coding for this program is contained in the Appendix.					
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CONTENTS

1	INTRODUCTION	1
2	USE OF COMPUTER SIMULATION	3
3	DEMONSTRATION OF THE PROBLEM	4
4	MANDEL'S SOLUTION	6
5	USE OF INTERACTIVE SOFTWARE PACKAGE	15
6	NEED FOR FUTURE RESEARCH	21
7	SUMMARY AND CONCLUSIONS	22
8	REFERENCES	23
9	APPENDICES	
	A Sensitivity Analyses	24
	B Fortran Coding For Mandel Subroutine	31

FIGURES

1	CONCEPTUAL ILLUSTRATION OF THE ORDINARY LEAST SQUARES TECHNIQUE	2
2	EFFECT OF SLOPE ON THE TRANSLATION OF X ERROR INTO APPARENT Y ERROR	7
3	CONCEPTUAL ILLUSTRATION OF MANDEL'S METHOD	9
4	COMPARISON OF DISTRIBUTIONS OF REGRESSION ESTIMATES	12
5	PRIMARY PRINTOUT OF X-ERROR SOFTWARE PACKAGE	16
6	TYPICAL REGRESSION RESULTS WITH CONCRETE STRENGTH DATA	18
7	PRINTOUT RESULTING FROM MENU ITEM 1	19
8	PRINTOUT RESULTING FROM MENU ITEM 2	20

TABLES

1	EXAMPLES OF BIAS INTRODUCED BY THE PRESENCE OF X ERROR	5
2	EFFECT OF SLOPE ON THE DEGREE OF BIAS INTRODUCED	8
3	COMPARISON OF MANDEL'S METHOD WITH ORDINARY LEAST SQUARES	11
4	EFFECT OF X ERROR ON INTERVAL ESTIMATES	14

1 INTRODUCTION

Many engineering applications require the development of a mathematical model (equation) to characterize some physical relationship. Examples include those such as the following:

<u>CHARACTERISTIC OF INTEREST</u>	<u>X DATA (Independent Variable)</u>	<u>Y DATA (Dependent Variable)</u>
Compressive strength of concrete	7-day test results	28-day test results
Rating of highway pavement serviceability	Output of mechanical roughness-measuring device	Average rating of a team of panelists
Rating of highway pavement serviceability	Cumulative axle loads	Current rating of serviceability

In the first example, the objective is a reliable early predictor of the 28-day strength of concrete, a measure upon which many acceptance procedures are based. The objective of the second example is to replace a costly and time consuming subjective rating procedure with a simple mechanical device. In the third example, a relationship is sought that will become an integral part of a pavement management system.

The variable to be predicted or estimated is placed on the Y axis and an equation of the form $Y = F(X)$ is desired. The equation may be linear, quadratic, exponential, or any other appropriate form. The analyst, from his understanding of the physical process, will often know the correct form in advance. In other cases, it may be necessary to let the data dictate the form.

The desired relationship is often derived empirically from a set of X, Y data values using the technique of least squares (1) as illustrated in Figure 1. The procedure, invisible to the analyst when executed by a

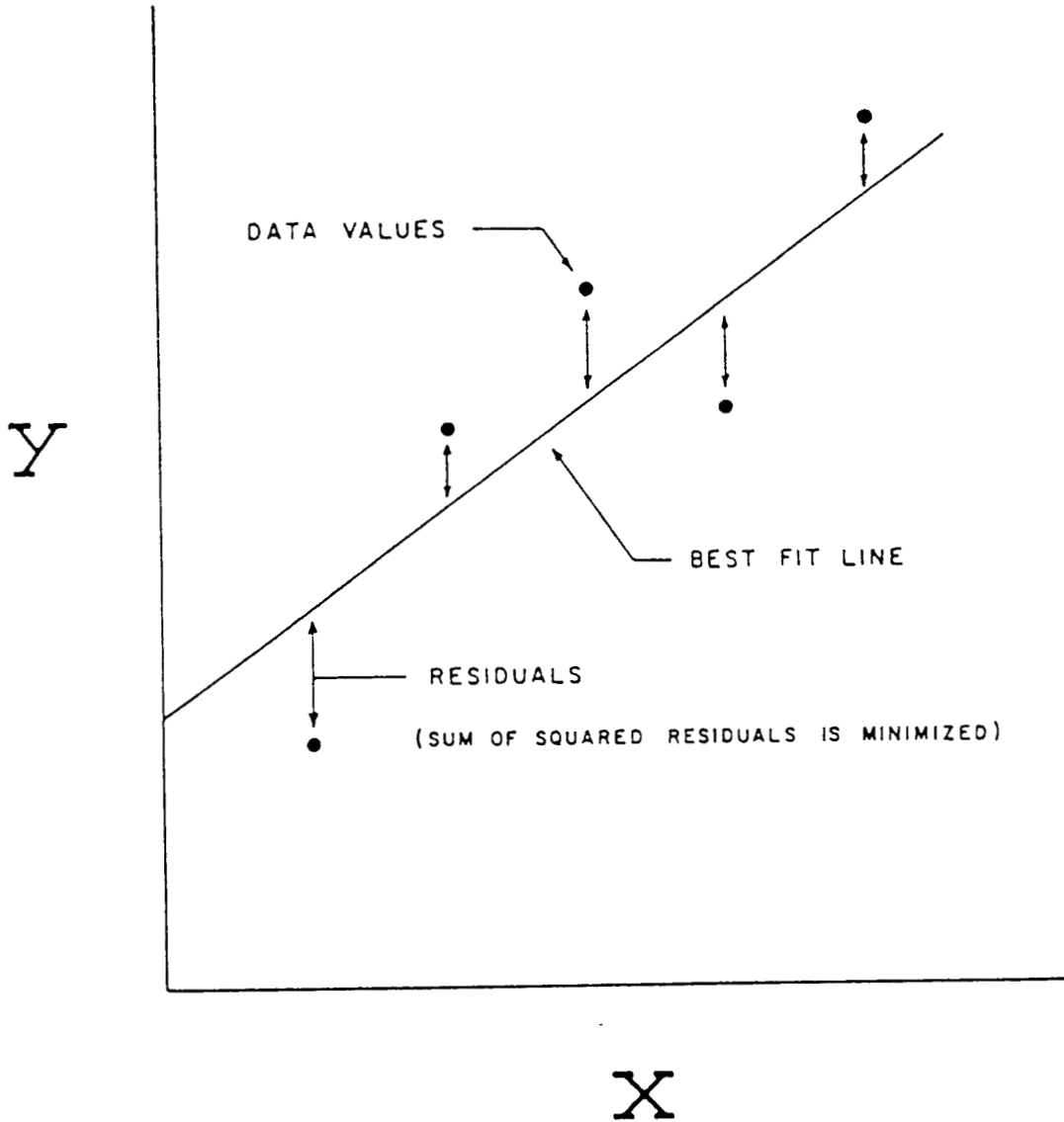


Figure 1. Conceptual illustration of the ordinary least squares technique.

computer program, consists of solving for the line that "best" fits the data. When ordinary least squares is used, the best fit is defined as that line of the chosen form that minimizes the sum of the squared residuals in a direction parallel to the Y axis.

In carrying out this procedure, several theoretical assumptions are made, one of the most fundamental of which is that the X values are known exactly and that any error of measurement is associated only with the Y values. Because it is often impossible or impractical to achieve this idealized condition in practice, this study was undertaken (a) to investigate the effect of failing to satisfy this assumption and (b) to develop and test a software package that incorporates a theoretical procedure for dealing with X error (2). This report addresses linear models since only the linear solution of the X-error problem has been published to date.

2 USE OF COMPUTER SIMULATION

In order to demonstrate the extent of the X-error problem and the effectiveness of the solution, a method was required to observe and quantify the accuracy and precision of the regression estimates. This can readily be accomplished with computer simulation by performing the following steps:

1. Randomly generate a bivariate normal X, Y data set with known regression (population) parameters:
 - (a) intercept (β_0)
 - (b) slope (β_1)
 - (c) residual error (σ_{yx})

2. Include a fixed amount of X error, either in absolute terms or as a percentage of σ_{yx} .
3. Use the randomly generated data to estimate the regression parameters:
 - (a) intercept (B_0)
 - (b) slope (B_1)
 - (c) residual error (S_{yx})
4. Repeat the entire process many times in order to compare the distributions of the regression estimates with the known parameters. Ideally, the sampling distributions of the estimates should be centered on the true population parameters and have relatively narrow dispersions.

This technique can be used to provide a very dramatic demonstration of the bias introduced by error in the X variable and the conditions that accentuate it. It will also be used to demonstrate the effectiveness of the procedure developed to overcome this problem.

3 DEMONSTRATION OF THE PROBLEM

Table 1 has been prepared using dimensionless data to illustrate the detrimental effect even a moderate amount of X error can have under certain conditions. It can be observed that, when there is no X error, the estimated values (averages for 1000 replications) of all three regression parameters are extremely close to the true population values. When the amount of X error is as little as 25 percent of the Y error (σ_{yx}), it can be seen that a substantial amount of bias has been introduced in the estimates of both the intercept and the residual error. As the degree of X error increases, all three regression parameters begin to show considerable

Table 1. Examples of bias introduced by the presence of X error.

PARAMETER	TRUE VALUE	REGRESSION ESTIMATES (a) OBTAINED BY ORDINARY LEAST SQUARES FOR SELECTED LEVELS OF X ERROR (b)				
		0	25	50	75	100
Intercept	100	100.14	108.11	129.25	160.81	200.23
Slope	10	10.00	9.84	9.41	8.78	8.00
Residual Error	5	4.93	13.21	24.62	35.08	44.79

(a) Results obtained by computer simulation with 1000 replications of 30 data points spanning the range between approximately $X = 30$ and $X = 70$.

(b) X error is measured as the ratio σ_{XX}/σ_{YX} , expressed as a percent, in which σ_{XX} and σ_{YX} represent the precision with which the X and Y variables are measured.

1
5
1

bias. When the X error and the Y error are approximately equal, a fairly common situation in actual practice, the regression estimates differ substantially from the true population parameters.

It should be noted that this example was chosen to dramatize the potentially serious nature of the X-error problem. Although the three population parameters ($\beta_0 = 100$, $\beta_1 = 10$, $\sigma_{yx} = 5$) are not extreme in any sense, the effect is pronounced because the slope is fairly steep. Figure 2 presents a conceptual illustration of the effect of the slope in translating X error into apparent Y error, error that the ordinary least squares procedure attributes solely to the Y variable. The examples in Table 2 further demonstrate this effect. Viewed collectively, the examples in Tables 1 and 2 provide empirical evidence of two conditions that may influence the magnitude of the X-error problem, i.e., the ratio of X error to Y error and the slope of the regression line.

4 MANDEL'S SOLUTION

A theoretically derived procedure for avoiding the bias in the regression estimates due to X error has been published by Mandel (2). Use of the procedure requires one additional bit of information that is usually readily available or readily obtainable: the ratio of the variances associated with the X and Y measurements. Although the mathematical procedure is somewhat involved, the concept is very easy to visualize. Ordinary least squares minimizes the sum of the squared residuals in a direction parallel to the Y axis. As shown in Figure 3, Mandel's procedure performs the minimization process in a direction oblique to the X and Y axes, the exact angle being determined primarily by the relative magnitude of the X and Y error.

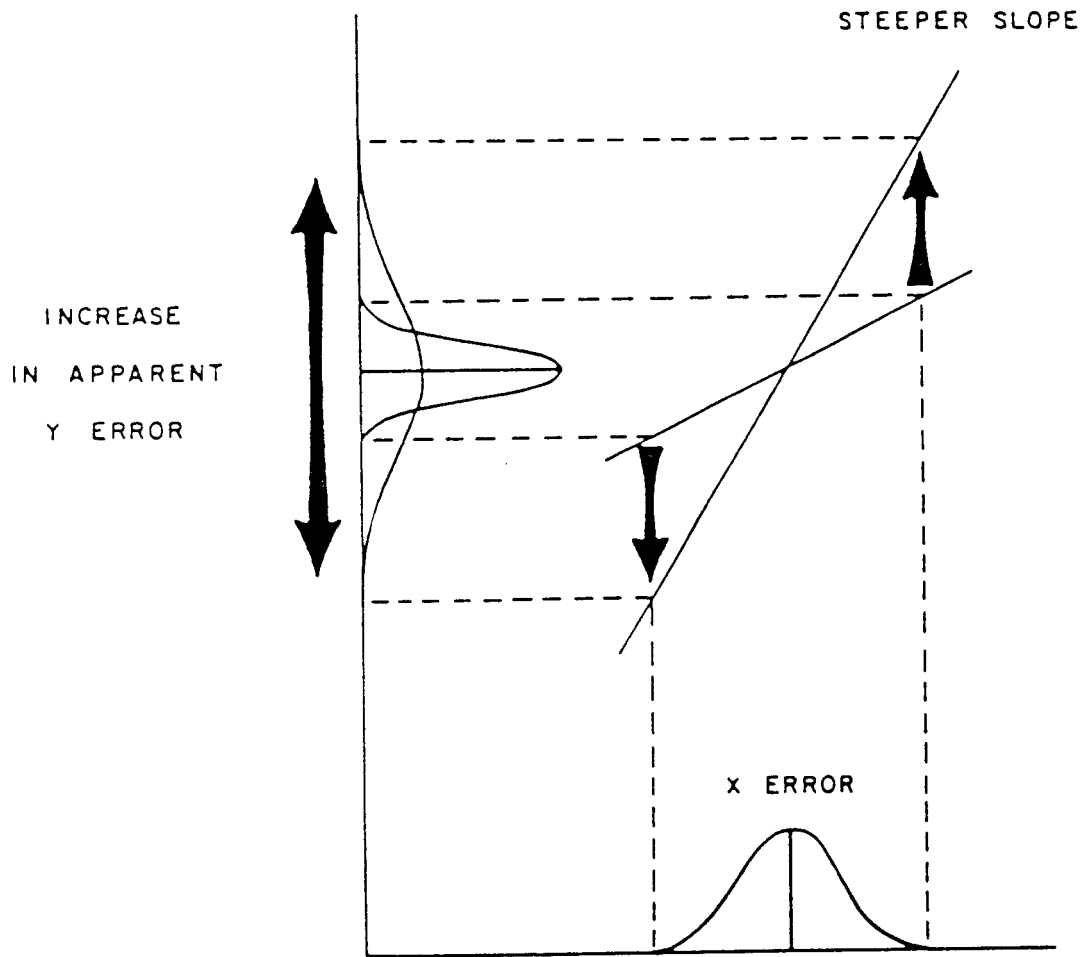


Figure 2. Effect of slope on the translation of X error into apparent Y error.

Table 2. Effect of slope on the degree of bias introduced.

PARAMETER	TRUE VALUE	REGRESSION ESTIMATES (a) OBTAINED BY ORDINARY LEAST SQUARES FOR FIXED X ERROR (b) AND SELECTED LEVELS OF SLOPE				
		0	25	50	75	100
Intercept	100	99.94	105.14	109.95	150.26	200.23
Slope	(c)	0.00	0.40	0.80	3.99	8.00
Residual Error	5	4.95	5.44	6.69	22.58	44.79

(a) Results obtained by computer simulation with 1000 replications of 30 data points spanning the range between approximately $X = 30$ and $X = 70$.

(b) The level of X error is fixed at 100 percent ($\sigma_{XX} = \sigma_{YX}$, where σ_{XX} and σ_{YX} represent the precision with which the X and Y variables are measured).

(c) Variable, values given in column headings.

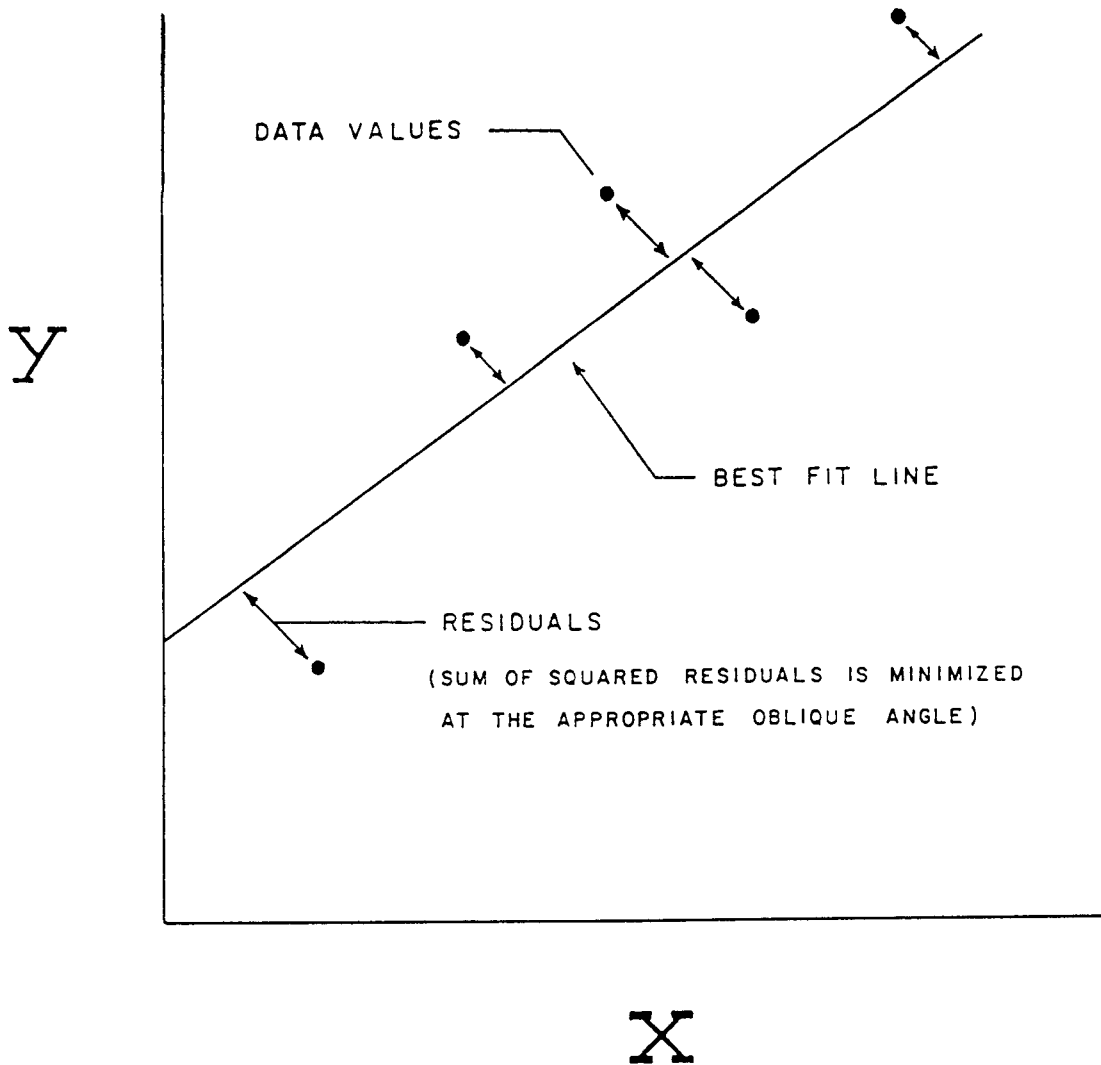


Figure 3. Conceptual illustration of Mandel's method.

In order to test the effectiveness of Mandel's method, it was applied to the same data sets used to develop Table 1. The results are reported in Table 3 and the values from Table 1 are repeated for ease of comparison. It can be seen from Table 3 that Mandel's method is extremely effective in removing the bias that exists when an application with X error is analyzed by ordinary least squares. Its only discernible weakness in this example is a possible small downward bias of the estimate of the intercept when the X error is quite large.

To judge whether or not this apparent bias was real, the simulation program was modified to print out a histogram and elementary statistics for 1000 intercept estimates. The X error was held constant at 100 percent of σ_{yx} . Although not strictly applicable because the distribution of intercept estimates was somewhat skewed, a t test indicated that the average intercept of 93.06 was highly significantly different ($\alpha < 0.001$) from the true value of 100.0. Although it is not obvious from the results in Table 3, a similar test suggests that the slope estimates may also be biased to a very small degree. Consequently, while Mandel's method appears to be very effective, and is far superior to ordinary least squares, it may not be not totally unbiased in all cases.

Figure 4 illustrates in a graphical way the effects that have been observed in Table 3. The distributions shown in this figure were drawn from histograms generated by the same simulation programs used to develop Tables 1 - 3. Mandel's method can be seen to be essentially unbiased in that the means of the distributions generated by that method are very close to the true population parameters. In marked contrast, the distributions produced by ordinary least squares are shifted substantially away from the true parameters. Another important observation in Figure 4 is that the

Table 3. Examples of bias introduced by the presence of X error.

PARAMETER	TRUE VALUE	METHOD	REGRESSION ESTIMATES (a) OBTAINED FOR SELECTED LEVELS OF X ERROR (b)				
			0	25	50	75	100
Intercept	100	Mandel	100.18	100.19	98.34	96.69	93.06
		OLS (c)	100.14	108.11	129.25	160.81	200.23
Slope	10	Mandel	10.00	9.99	10.03	10.06	10.14
		OLS (c)	10.00	9.84	9.41	8.78	8.00
Residual Error	5	Mandel	4.93	4.94	4.97	4.94	4.97
		OLS (c)	4.93	13.21	24.62	35.08	44.79

(a) Results obtained by computer simulation with 1000 replications of 30 data points spanning the range between approximately $X = 30$ and $X = 70$.

(b) X error is measured as the ratio σ_{XX}/σ_{YY} , expressed as a percent, in which σ_{XX} and σ_{YY} represent the precision with which the X and Y variables are measured.

(c) Ordinary least squares.

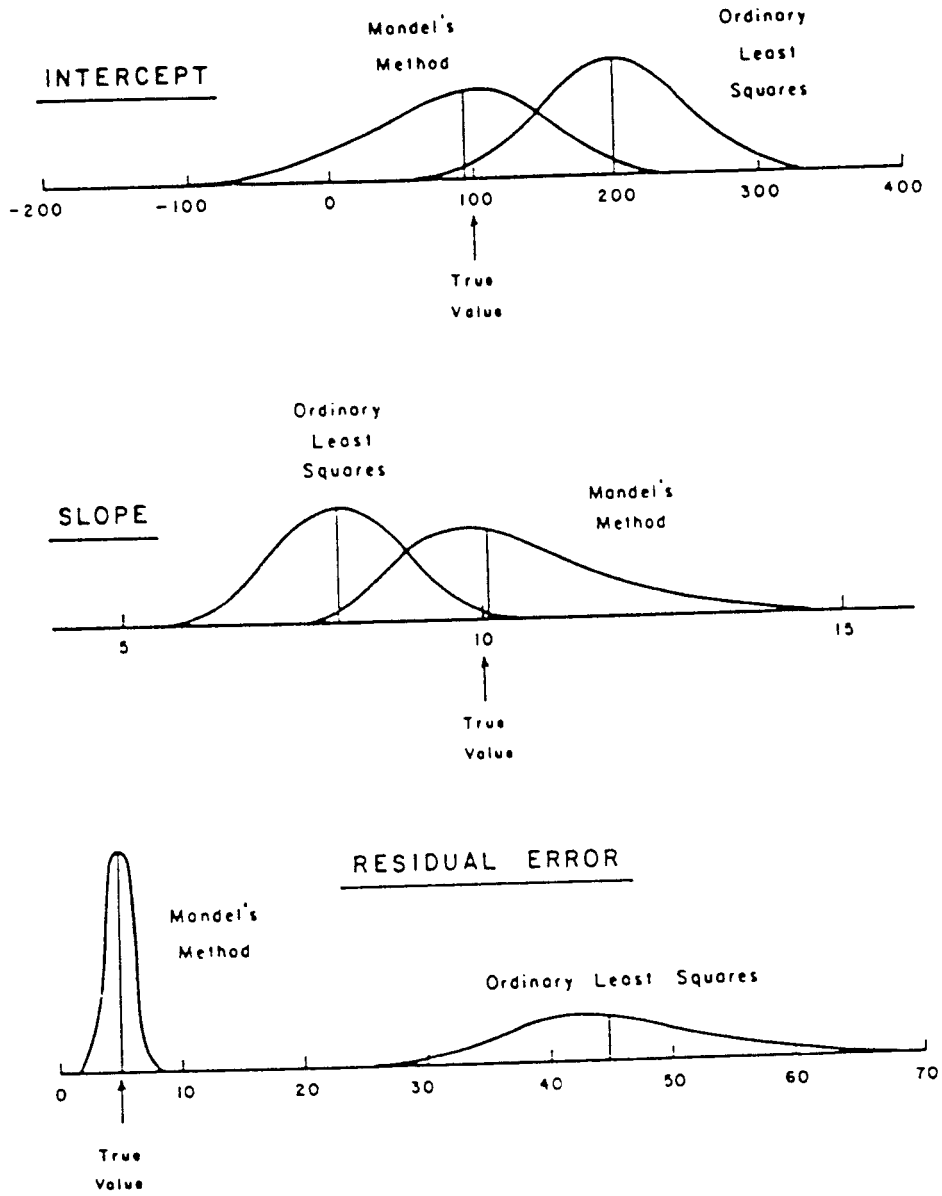


Figure 4. Comparison of distributions of regression estimates.

distributions for the intercept and the slope obtained with Mandel's method are only slightly more dispersed than those obtained by ordinary least squares. This indicates that a substantial gain in accuracy has been achieved with only a slight loss of precision. For the residual error, Mandel's method is both more accurate and more precise.

An interesting feature of Mandel's method is that, unlike the least squares technique, the same regression line will be obtained regardless of which variable is considered to be independent (X) and which is dependent (Y). Furthermore, the degree of uncertainty associated with predictions made by this method is the same for either choice of variables (2, p. 9). This property conveniently avoids a controversial aspect of the calibration application, the need to work backward through the regression procedure to estimate what value of X gave rise to an observed value of Y.

Another series of computer simulation tests was performed using the standard procedures for computing interval estimates for the intercept, slope, and σ_{yx} . A level of confidence of $1 - \alpha = 0.95$ was selected and the number of times the interval estimate actually contained the true population parameter was counted. For 1000 replications, the empirically observed results should fall within the range of approximately $0.95 \pm 2(((0.95)(0.05))/1000)^{1/2} = 0.95 \pm 0.014$ when the interval estimation process is working properly. It can be seen from the results in Table 4 that, even for small amounts of X error, the interval estimates computed by ordinary least squares contain the population parameters substantially less often than desired. In contrast, all of the interval estimates computed by Mandel's method are satisfactory even though the standard confidence interval estimation procedures are known to be approximate.

Table 4. Effect of X error on interval estimates.

PARAMETER	DESIRED CONFIDENCE LEVEL	METHOD	EMPIRICALLY OBSERVED (a) CONFIDENCE LEVELS AT SELECTED LEVELS OF X ERROR (b)				
			0	25	50	75	100
Intercept ($\beta_0=100$)	0.95	Mandel	0.950	0.938	0.954	0.939	0.953
		OLS (c)	0.951	0.898	0.768	0.548	0.321
Slope ($\beta_1=10$)	0.95	Mandel	0.948	0.941	0.956	0.948	0.952
		OLS (c)	0.946	0.889	0.755	0.534	0.301
Residual Error ($\sigma_{YX}=5$)	0.95	Mandel	0.955	0.956	0.963	0.947	0.959
		OLS (c)	0.955	0.0	0.0	0.0	0.0

(a) Results obtained by computer simulation with 1000 replications of 30 data points spanning the range between approximately $X = 30$ and $X = 70$.

(b) X error is measured as the ratio σ_{XX}/σ_{YX} , expressed as a percent, in which σ_{XX} and σ_{YX} represent the precision with which the X and Y variables are measured.

(c) Ordinary least squares.

5 USE OF INTERACTIVE SOFTWARE PACKAGE

The following example illustrates the use of the Fortran program developed to make Mandel's solution widely available to a broad range of potential users. (A complete listing of the Fortran program is contained in Appendix B.) It is based on concrete strength data collected from a construction project in New Jersey and is a contrived example in that the data set that is used was randomly generated from a population having the same statistical parameters as those observed in the field. This approach provides a known control against which the results obtained by the two methods may be compared. Otherwise, it could only be observed that the results obtained by the two methods were distinctly different and it would not be known how close either one came to estimating the true population parameters.

In order to use this program, two predetermined values must be entered. These are (a) the ratio of the X and Y error variances and (b) the degree of correlation between the X and Y errors. Note that this latter value represents the correlation between the X and Y errors, not the correlation between the X and Y measurements. Except for certain specialized applications (2, p.3), this correlation is normally zero. The ratio of error variances has been set at 1.0 because, at the relatively high levels of 7-day and 28-day strengths used in this example, it is believed that the measurement error is essentially the same for both sets of data.

Figure 5 presents the primary portion of the printout for the X-error software package. In addition to the two predetermined values, the program prints elementary statistics, basic regression parameters, and the approximate significance levels with which the intercept and slope may be

LINEAR REGRESSION USING MANDEL PROCEDURE FOR ERROR IN THE INDEPENDENT VARIABLE
 =====

PREDETERMINED RATIO VAR(X-ERROR)/VAR(Y-ERROR) = 1.000

PREDETERMINED CORRELATION OF X-ERROR WITH Y-ERROR = .000

30 DATA POINTS

	X	Y
MINIMUM	2400.000000	3590.000000
MAXIMUM	4440.000000	5590.000000
MEAN	3376.333252	4634.000000
STANDARD DEVIATION	535.559326	520.706299
SKEW	.12	.06
KURTOSIS	-.69	-.77

PARAMETER ESTIMATES

	Y = F(X)	X = F(Y)
INTERCEPT	1389.104635	-1445.371808
SLOPE	.961071	1.040506
RESIDUAL ERROR, S(YX) AND S(XY)	290.212512	290.212512
MANDEL'S BK STATISTIC	.92	1.08
STANDARD ERROR OF INTERCEPT	512.283101	757.217146
STANDARD ERROR OF SLOPE	.150158	.162569

APPROXIMATE SIGNIFICANCE LEVELS

NULL HYPOTHESIS	Y = F(X)	X = F(Y)
INTERCEPT IS ZERO	.006	.033
SLOPE IS ZERO	.000	.000

SELECT ONE OF THE FOLLOWING OPTIONS

1. INTERVAL ESTIMATES FOR INTERCEPT, SLOPE, AND RESIDUAL ERROR
2. INTERVAL ESTIMATES FOR TRUE Y AT SELECTED VALUE OF X
3. INTERVAL ESTIMATES FOR TRUE X AT SELECTED VALUE OF Y
4. TERMINATE THIS RUN

?

Figure 5. Primary printout of X-error software package.

inferred to be nonzero. The data is treated as both $Y = F(X)$ and $X = F(Y)$ in the primary printout and also in the optional portions that follow.

As noted in the preceding section, Mandel's procedure produces the same regression line regardless of which variable is to be predicted. This can readily be demonstrated from this printout, for example, by solving for X in the equation $Y = F(X)$ and comparing with the equation $X = F(Y)$ obtained from the same printout.

The data points and regression results are shown in Figure 6. The estimated slope and intercept are typical in that they are examples of central values of the distributions shown in Figure 4. Like the dimensionless examples in Table 4, ordinary least squares has produced a considerably biased estimate while Mandel's method has produced an estimate very close to the true location of the line.

Figure 5 also shows the menu that is printed following the primary output. This enables the user to compute various interval estimates or, if these are not desired, to terminate the run.

If Item 1 is selected from the menu, the output shown in Figure 7 is obtained. This provides both approximate single-sided and approximate double-sided interval estimates at several commonly used significance levels for the three basic parameters of linear regression -- intercept, slope, and residual error.

Figure 8 shows the output resulting from the selection of Item 2 from the menu. In this case, the user is first required to enter the value of X at which interval estimates for Y are to be computed. The program then prints out a series of approximate single-sided and double-sided limits, exactly as was done for Item 1. The procedure for Item 3 is the same

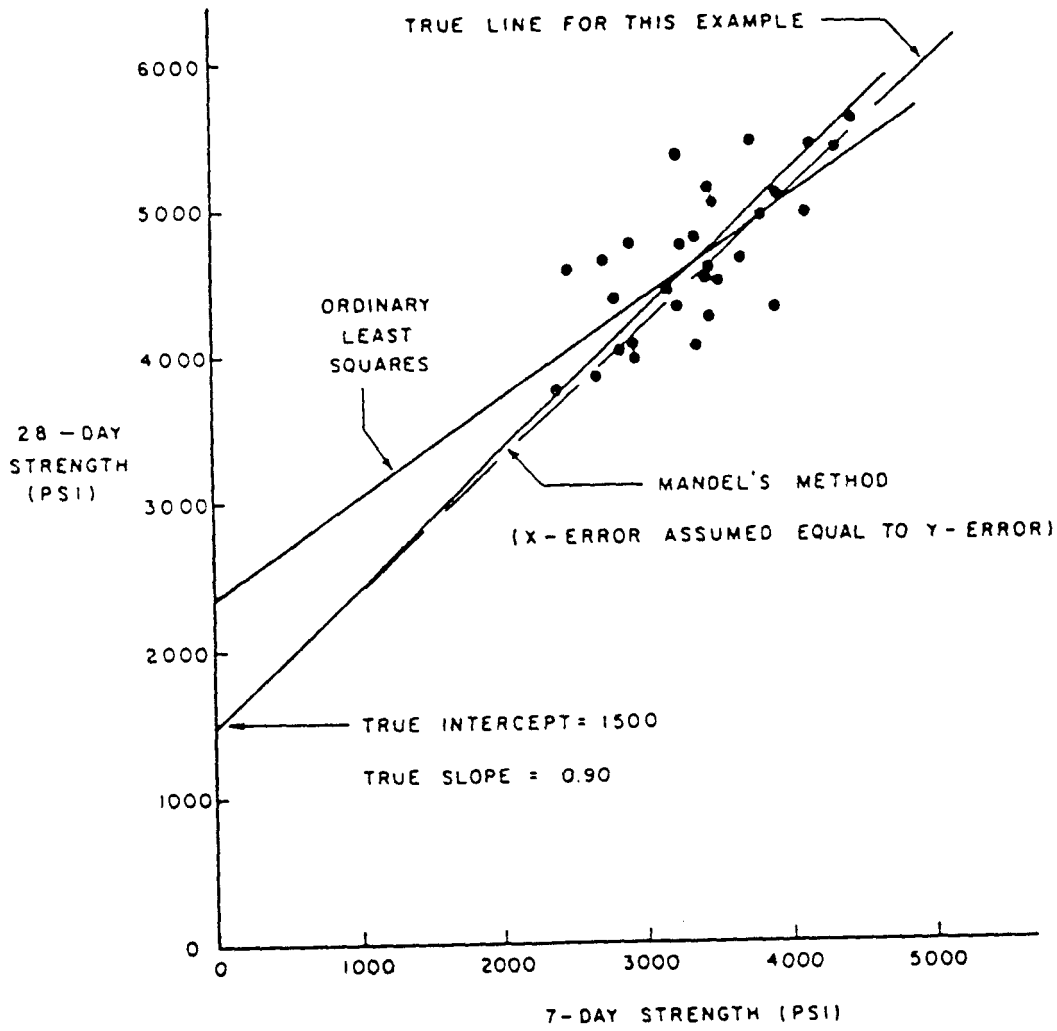


Figure 6. Typical regression results with concrete strength data.

APPROXIMATE INTERVAL ESTIMATES FOR INTERCEPT

TAIL AREA	Y = F(X)	X = F(Y)
LOWER .005	-26.553672	-3537.888114
LOWER .010	125.189853	-3313.592605
LOWER .025	339.721983	-2996.487849
LOWER .050	517.637687	-2733.506656
LOWER .100	716.716942	-2439.243143
UPPER .100	2061.492327	-451.500473
UPPER .050	2260.571582	-157.236960
UPPER .025	2438.487286	105.744233
UPPER .010	2653.019416	422.848989
UPPER .005	2804.762941	647.144498

APPROXIMATE INTERVAL ESTIMATES FOR SLOPE

TAIL AREA	Y = F(X)	X = F(Y)
LOWER .005	.546119	.591257
LOWER .010	.590597	.639412
LOWER .025	.653480	.707492
LOWER .050	.705630	.763952
LOWER .100	.763983	.827129
UPPER .100	1.158158	1.253883
UPPER .050	1.216512	1.317060
UPPER .025	1.268662	1.373520
UPPER .010	1.331544	1.441600
UPPER .005	1.376023	1.489755

APPROXIMATE INTERVAL ESTIMATES FOR RESIDUAL ERROR

TAIL AREA	S(YX)	S(XY)
LOWER .005	215.050256	215.050256
LOWER .010	221.014407	221.014407
LOWER .025	230.306780	230.306780
LOWER .050	238.849756	238.849756
LOWER .100	249.393031	249.393031
UPPER .100	352.869141	352.869300
UPPER .050	373.245117	373.245152
UPPER .025	392.498291	392.498435
UPPER .010	416.953857	416.953873
UPPER .005	435.017822	435.017952

SELECT ONE OF THE FOLLOWING OPTIONS

1. INTERVAL ESTIMATES FOR INTERCEPT, SLOPE, AND RESIDUAL ERROR
2. INTERVAL ESTIMATES FOR TRUE Y AT A SELECTED VALUE OF X
3. INTERVAL ESTIMATES FOR TRUE X AT A SELECTED VALUE OF Y
4. TERMINATE THIS RUN

?

Figure 7. Printout resulting from Menu Item 1.

ENTER VALUE OF X AT WHICH COMPUTATIONS ARE MADE
?
3000

POINT ESTIMATE OF TRUE Y = 4272.317024

S(Y FOR OBSERVED X) = 293.917236

APPROXIMATE INTERVAL ESTIMATES

TAIL AREA	Y FOR OBSERVED XDEL
LOWER .005	3460.105551
LOWER .010	3547.162096
LOWER .025	3670.250253
LOWER .050	3772.321978
LOWER .100	3886.541402
UPPER .100	4658.092645
UPPER .050	4772.311789
UPPER .025	4874.389681
UPPER .010	4997.475315
UPPER .005	5084.536625

SELECT ONE OF THE FOLLOWING OPTIONS

1. INTERVAL ESTIMATES FOR INTERCEPT, SLOPE, AND RESIDUAL ERROR
2. INTERVAL ESTIMATES FOR TRUE Y AT SELECTED VALUE OF X
3. INTERVAL ESTIMATES FOR TRUE X AT SELECTED VALUE OF Y
4. TERMINATE THIS RUN

?

Figure 8. Printout resulting from Menu Item 2.

except that, in this case, a value of Y must be entered in order to obtain approximate interval estimates for X.

The entire menu appears again after each selection is acted upon. In this way, the user may repeat any selection or, by selecting Item 4, the interactive session may be terminated.

6 NEED FOR FUTURE RESEARCH

This report has dealt entirely with linear regression applications because theoretical solutions for the more complex models apparently have not been derived or published to date. Obvious extensions of this methodology that would be immediately useful are the following:

1. Regression line forced to go through the origin.
2. Multiple linear regression and other curvilinear models.
3. Nonlinear regression models.

It was shown that results obtained by linear regression can be greatly in error when X error is present. It is reasonable to believe that similar problems exist with the more complex models. Since many applications require the line to go through the origin, or take some type of curvilinear form, it would seem highly desirable to develop these more advanced techniques in a manner that can properly account for the presence of X error.

Ideally, if the appropriate talents are brought to bear upon this problem, it is believed that a theoretical solution can be found as it was for the linear case. If for some reason a theoretical solution is not obtainable, it may still be possible to empirically derive a practical solution that can be shown to substantially improve upon the estimates that are made.

7 SUMMARY AND CONCLUSIONS

Regression analysis is frequently used in the engineering profession to develop mathematical models for many different applications. Regression theory assumes that the X values are known without error, a requirement that often cannot be met. Computer simulation was used to demonstrate that, under certain conditions, regression estimates obtained by ordinary least squares can be seriously in error. A quantitative measure of the severity of the problem is discussed in Appendix A.

The consequences of this finding may be alarming. In the presence of X error, OLS estimates may often be unreliable. The degree to which this ultimately affects the conclusions of research studies or influences policy decisions is not known, but the potential harm of specifying the wrong material or product, or of establishing a less-than-optimal policy or design, is recognized to be substantial. An error of this type will seldom be an isolated case; it will be repeated with each subsequent application of the product or design standard.

It was also demonstrated by computer simulation that Mandel's method is extremely effective at removing most of the bias introduced by error in the X variable. Figure 4 and Tables 3 and 4 clearly show that, in general, Mandel's method provides substantially more accurate results than ordinary least squares and Figure 6 illustrates this fact with a specific example based on concrete strength tests. The complete theoretical development is contained in the source document (2).

To make this technique widely available to a broad range of users, Appendix B provides the Fortran coding for an extremely user-friendly interactive program. Unlike many commercial software packages, the needs of the typical user have been anticipated to the extent that a wide variety

of statistical estimates may be obtained directly from the printout without the need for external tables or specialized expertise.

To develop the full range of this important new technique, additional research will be required. Desirable extensions of this work include regression lines forced to go through the origin, various curvilinear forms, multiple linear regression, and X-error procedures coupled with nonlinear regression techniques.

8 REFERENCES

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2. J. Mandel. Fitting Straight Lines when Both Variables are Subject to Error. Journal of Quality Technology, Volume 16, No. 1, pp. 1-14, 1984.

Appendix A: SENSITIVITY ANALYSES

Computer simulation analyses were performed to determine how robust the Mandel regression procedure was with respect to: (1) mis-specified lambda (i.e. the ratio of Y error to X error variances), (2) slope variations, and (3) various ranges of X and Y observations. In this simulation the true equation subject to estimation was known, allowing the effectiveness of the Mandel procedure to be gaged by monitoring the ratios of the mean parameter estimates to their true values. A ratio of 1.0 for the slope parameter would indicate that the average estimated slope was correctly estimated as equal to the true value. A ratio of 0.5, on the other hand, would indicate that the average estimated slope was biased downwards by 50 per cent.

This sensitivity analysis required that the Mandel procedure be tested throughout the range of possible regression applications. This meant more than simply testing small and large numbers. Linear scaling of the X and Y observations may superficially change the apparent slope of the reported regression line but, in fact, linear scaling has no effect on the actual regression analysis itself other than cosmetic representation. Consequently a more meaningful criteria was developed to test the Mandel procedure.

The key to this testing procedure lay in the development of two ratios: that of the range of X observations to the residual error in X and that of the range of Y observations to the residual error in Y. The significance of these ratios may be readily appreciated by the schematic diagram presented in Figure A1. Consider the hypothetical case in which X error is three times as large as the error in Y. An excellent estimate of

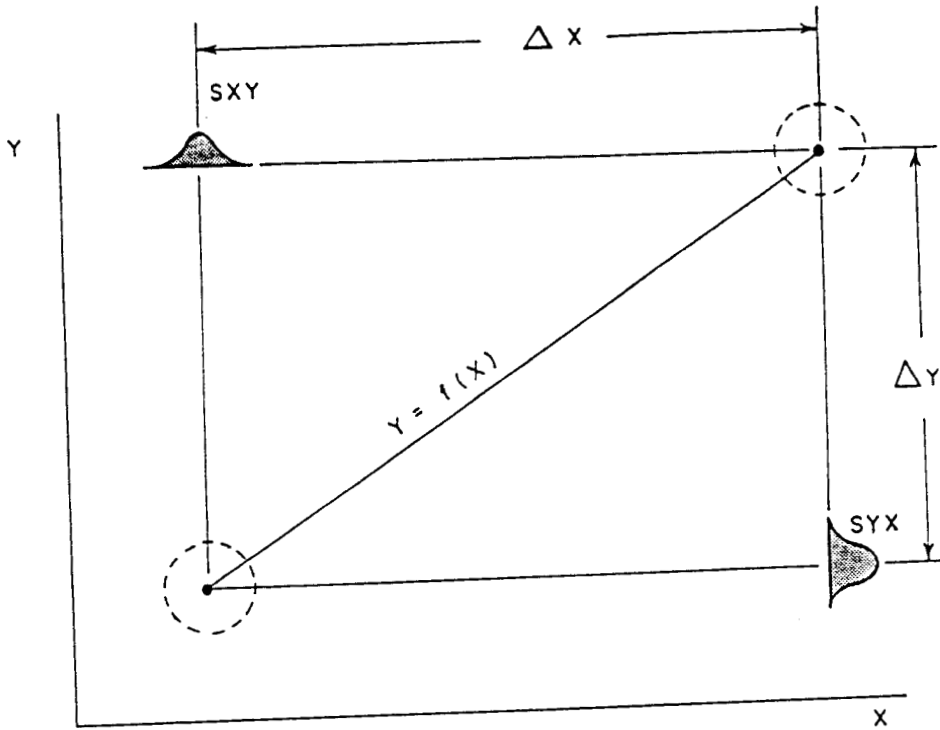


Figure A.1. Schematic representation of the influence the ratios $(\Delta X/S_{XY})$ and $(\Delta Y/S_{YX})$ have on the estimation of $Y=f(X)$.

the $Y=f(X)$ relationship may be obtained, even with OLS and regardless of the true slope, provided the observed data spans a range substantially broader than the potential error in the X values. That is, provided the locus of (X,Y) coordinates at one extreme end of the $Y=f(X)$ relationship is substantially distinct from the locus of (X,Y) coordinates at the other extreme end, the line passing through the center of the two data masses will be easily identifiable and more precisely estimated as the two loci are separated.

The ratios $(\Delta X)/(S_X Y)$ and $(\Delta Y)/(S_Y X)$ describe the degree to which these loci are offset from one another. Ratios of 100 and 100 would be associated with relationships so strong that they may be well estimated graphically. Ratios of 5 and 5 (or lower) would be associated with visually more obscure trends. Of course, ratios of 5 and 100 or any other magnitude and combination may actually occur and can as easily be checked.

The equation first presented in Table 1 of this report, $Y = 100.0 + 10.0 \cdot X$, was used as the nominal point of departure for this sensitivity analysis. Lambda was first correctly specified, then over and under estimated by a factor of two, and then over estimated by a factor of ten thousand (essentially ignoring X error and approximating an OLS analysis). One thousand data sets each for the sample sizes of 10, 30 and 100 (X,Y) coordinate pairs were generated under a variety of $(\Delta X)/(S_X Y)$ and $(\Delta Y)/(S_Y X)$ conditions. The Mandel regression procedure was applied to each data set, and the accuracy of the parameter estimates was monitored as previously described. Additionally, a tally was kept to empirically determine the frequency with which the true parameter values fell outside their estimated 95 percent confidence limits. Then, for even more extreme conditions, the entire sensitivity analysis was reproduced with the slope

both increased and decreased fifty-fold, using the true relationships of $Y = 100 + 500.0 \cdot X$ and $Y = 100 + 0.2 \cdot X$, respectively. Partial results from these investigations are presented in Tables A1 and A2.

Several general conclusions may be drawn from this sensitivity analysis. They are:

- 1) Properly used, the Mandel procedure is accurate and generally reliable regardless of the data's variability.
- 2) Proper use of the Mandel procedure produces significance levels quite close to the nominal level specified. Apparently the Student t distribution used for this approximation is very close to the as-yet-unknown statistical distribution appropriate for this step.
- 3) The intercept appears to be the most sensitive of all estimated parameters, especially when the origin lies substantially outside the data range. This is intuitively an expected result, however, as in such cases the intercept is an extrapolated value. Thus this sensitivity is not considered to represent a significant shortcoming.
- 4) Mis-specification of lambda has relatively little effect in degrading the accuracy of slope parameter estimates, especially when the data spread is large relative to the inherent error. The intercept estimates are somewhat more affected, especially when the origin lies well outside the range of the data.
- 5) Mis-specification of lambda necessarily biases at least one of the error estimates - - either SYX or SXY, or both. This is mathematically unavoidable.

Table A1. Sensitivity of Mandel's mean parameter estimates.

PARAMETER	ΔX/SX	ΔY/SY	n	(MEAN PARAMETER ESTIMATE) / (TRUE PARAMETER VALUE)														
				Y = 100.0 + 10.0*X, SYX=5.0				Y = 100.0 + 500.0*X, SYX=1.0				Y = 100.0 + 0.2*X, SYX=1.0						
				TRUE/ESTIMATED LAMBDA			0.0001	TRUE/ESTIMATED LAMBDA			0.5	0.0001	TRUE/ESTIMATED LAMBDA			1.0	2.0	0.5
INTERCEPT	10	100	30	0.99	0.97	1.00	1.47	0.52	1.32	0.94	1.40	1.00	1.00	1.00	1.00	1.00	1.00	
	10	100	100	0.99	0.99	1.00	1.53	0.74	0.83	1.28	1.24	1.00	1.00	1.00	1.00	1.00	1.00	
	10	100	30	0.95	0.49	1.34	1.91	-1.62	-28.97	19.81	46.57	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	5	5	100	0.99	0.46	1.42	1.98	-0.44	-30.08	24.34	54.37	1.00	1.00	1.00	1.00	1.00	1.00	1.00
SLOPE	10	100	30	1.00	1.01	1.00	0.91	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	10	100	100	1.00	1.00	1.00	0.90	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	10	100	30	1.02	1.16	0.89	0.71	1.02	1.20	0.87	0.70	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	5	5	100	1.00	1.17	0.86	0.68	1.01	1.21	0.84	0.64	1.00	1.00	1.00	1.00	1.00	1.00	1.00
SYX ²	10	100	30	1.00	0.50	1.98	84.13	1.01	0.98	1.01	1.01	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	10	100	100	1.00	0.50	1.97	84.07	1.01	0.99	1.01	1.01	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	10	100	30	0.99	0.66	1.28	1.70	0.98	0.65	1.30	1.69	1.00	1.00	1.00	1.00	1.00	1.00	1.00
	5	5	100	1.00	0.65	1.30	1.68	1.00	0.64	1.29	1.64	1.00	1.00	1.00	1.00	1.00	1.00	1.00
SXY ²	10	100	30	1.00	1.00	0.99	0.08	1.01	1.96	0.50	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	10	100	100	1.00	1.01	0.99	0.00	1.01	1.97	0.50	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	10	100	30	0.99	1.32	0.64	0.00	0.98	1.30	0.65	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	5	5	100	1.00	1.29	0.65	0.00	1.00	1.28	0.64	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00

Notes: 1) Computer simulation used to perform sensitivity analysis. Values in body of table represent the average of 1000 replicate runs. Values of 1.0 indicate no apparent bias.
 2) LAMBDA = (SYX/SXY)²
 3) ΔX/SX = (Range of true X values)/SYX
 4) ΔY/SY = (Range of true Y values)/SYX

Table A2. Empirical significance levels for Mandel's parameter estimates.

PARAMETER		ΔX/SX		ΔY/SY		n		EMPIRICAL SIGNIFICANCE LEVELS, 0.05 ERROR RATE DESIRED											
								Y = 100.0 + 10.0*X, SYX=5.0				Y = 100.0 + 500.0*X, SYX=1.0				Y = 100.0 + 0.2*X, SYX=1.0			
								TRUE/ESTIMATED LAMBDA		TRUE/ESTIMATED LAMBDA		TRUE/ESTIMATED LAMBDA		TRUE/ESTIMATED LAMBDA		TRUE/ESTIMATED LAMBDA		TRUE/ESTIMATED LAMBDA	
		1.0	2.0	0.5	0.0001	1.0	2.0	0.5	0.0001	1.0	2.0	0.5	0.0001	1.0	2.0	0.5	0.0001		
INTERCEPT	10	0.04	0.06	0.06	0.31	0.04	0.05	0.04	0.04	0.04	0.07	0.05	0.04	0.04	0.07	0.05	0.04		
	100	0.05	0.05	0.05	0.84	0.04	0.06	0.05	0.05	0.05	0.06	0.05	0.05	0.05	0.06	0.05	0.05		
	100	0.09	0.07	0.20	0.55	0.10	0.07	0.22	0.57	0.11	0.07	0.22	0.57	0.11	0.07	0.20	0.58		
	5	0.11	0.30	0.45	0.99	0.12	0.36	0.47	0.99	0.09	0.30	0.42	0.99	0.09	0.30	0.42	0.99		
	5	0.04	0.06	0.05	0.34	0.04	0.05	0.04	0.04	0.04	0.04	0.07	0.06	0.04	0.07	0.06	0.04		
SLOPE	10	0.05	0.05	0.04	0.87	0.04	0.06	0.05	0.05	0.05	0.06	0.05	0.05	0.05	0.06	0.05	0.05		
	100	0.09	0.07	0.20	0.56	0.10	0.07	0.22	0.57	0.11	0.07	0.22	0.57	0.11	0.07	0.20	0.59		
	100	0.09	0.07	0.20	0.56	0.10	0.07	0.22	0.57	0.11	0.07	0.22	0.57	0.11	0.07	0.20	0.59		
	5	0.10	0.30	0.46	0.99	0.12	0.36	0.47	0.99	0.09	0.30	0.42	0.99	0.09	0.30	0.42	0.99		
	5	0.05	0.66	0.74	1.00	0.05	0.05	0.04	0.05	0.05	0.04	0.04	0.05	0.05	0.04	0.04	0.05		
SYX ²	10	0.05	1.00	1.00	1.00	0.05	0.05	0.06	0.06	0.05	0.04	0.06	0.06	0.04	0.04	0.04	0.06		
	100	0.05	0.28	0.19	0.57	0.06	0.29	0.19	0.55	0.04	0.31	0.18	0.57	0.04	0.31	0.18	0.57		
	100	0.05	0.28	0.19	0.57	0.06	0.29	0.19	0.55	0.04	0.31	0.18	0.57	0.04	0.31	0.18	0.57		
	5	0.05	0.86	0.49	0.96	0.05	0.85	0.44	0.95	0.03	0.82	0.49	0.96	0.03	0.82	0.49	0.96		
	5	0.05	0.05	0.06	1.00	0.05	0.76	0.66	1.00	0.05	0.75	0.66	1.00	0.05	0.75	0.66	1.00		
SXY ²	10	0.05	0.06	0.05	1.00	0.05	1.00	1.00	1.00	0.04	1.00	1.00	1.00	0.04	1.00	1.00	1.00		
	100	0.05	0.21	0.35	1.00	0.06	0.20	0.31	1.00	0.04	0.21	0.31	1.00	0.04	0.21	0.31	1.00		
	100	0.05	0.21	0.35	1.00	0.06	0.20	0.31	1.00	0.04	0.21	0.31	1.00	0.04	0.21	0.31	1.00		
	5	0.05	0.48	0.83	1.00	0.05	0.44	0.86	1.00	0.03	0.49	0.82	1.00	0.03	0.49	0.82	1.00		
	5	0.05	0.05	0.06	1.00	0.05	0.76	0.66	1.00	0.05	0.75	0.66	1.00	0.05	0.75	0.66	1.00		

Notes: 1) Computer simulation used to perform sensitivity analysis. Values in body of table represent the error proportion of 1000 replicate runs. Values of 0.05 indicate nominal risk level achieved.
 2) LAMBDA = (SYX/SXY)²
 3) ΔX/SX = (Range of true X values)/SXY
 4) ΔY/SY = (Range of true Y values)/SYX

- 6) Parameter estimates using OLS when the Mandel procedure is appropriate (i.e., extreme mis-specification of lambda,) may be generally tolerable provided the data spread is large relative to the inherent variability.
- 7) Inferences about estimated parameters become progressively less reliable as the degree to which lambda is mis-specified increases. In the most extreme case, where the procedure essentially defaults to OLS, not only are the nominal risk levels understated but they become progressively worse as the sample size is increased.

In summary, these sensitivity tests tend to confirm that, properly used, the Mandel regression procedure is an accurate, precise, and powerful tool over a very broad range of potential applications. While it can hardly be recommended that any tool be misused, it also appears that this particular regression procedure is moderately forgiving of inadvertent abuse. Thus it is highly recommended for immediate implementation wherever appropriate.

```

C =====MAN00010
C SUBROUTINE MANDEL(N,X,Y,VXY,RHO,PRINT)MAN00020
C MAN00030
C MAN00040
C MAN00050
C MAN00060
C ACKNOWLEDGEMENTMAN00070
C -----MAN00080
C THEORETICAL DEVELOPMENT -MAN00090
C DR. JOHN MANDEL, NATIONAL BUREAU OF STANDARDSMAN00100
C MAN00110
C FORTRAN CODING -MAN00120
C RICARDO BARROS, WITH THE ASSISTANCE OF RICHARD WEED,MAN00130
C BOTH OF THE NEW JERSEY DEPT. OF TRANSPORTATION.MAN00140
C AUGUST 22, 1988.MAN00150
C MAN00160
C MAN00170
C MAN00180
C SUMMARYMAN00190
C -----MAN00200
C THIS SUBROUTINE ESTIMATES AND, OPTIONALLY, REPORTS THEMAN00210
C RELEVANT PARAMETERS OF THE REGRESION LINES Y=F(X) AND X=F(Y)MAN00220
C WHEN BOTH X AND Y VARIABLES ARE SUBJECT TO MEASUREMENT ERROR.MAN00230
C MAN00240
C AS THE X ERROR APPROACHES ZERO, THE ESTIMATED INTERCEPT ANDMAN00250
C SLOPE PARAMETERS PRODUCED BY MANDEL APPROACH THE VALUESMAN00260
C PRODUCED BY CONVENTIONAL ORDINARY LEAST SQUARES TECHNIQUES.MAN00270
C MAN00280
C THE SOURCE DOCUMENT FOR THE UNDERLYING THEORY IS 'FITTINGMAN00290
C STRAIGHT LINES WHEN BOTH VARIABLES ARE SUBJECT TO ERROR' BYMAN00300
C DR. JOHN MANDEL AND PUBLISHED IN THE JOURNAL OF QUALITYMAN00310
C TECHNOLOGY, VOL. 16, NO. 1, JANUARY 1984.MAN00320
C MAN00330
C FURTHER DISCUSSION OF THE X ERROR PROBLEM IS PRESENTED INMAN00340
C 'APPLIED REGRESSION IN THE PRESENCE OF X-ERROR'MAN00350
C BY RICARDO BARROS AND RICHARD WEED OF THE NEW JERSEYMAN00360
C DEPARTMENT OF TRANSPORTATION.MAN00370
C MAN00380
C THIS SUBROUTINE REQUIRES LESS THAN 0.2 SECONDS TO ESTIMATEMAN00390
C REGRESSION PARAMETERS FOR AN ARRAY OF 1000 (X,Y) DATA
C OBSERVATIONS ON AN IBM 3081.MAN00400
C MAN00410
C MAN00420
C MAN00430
C MAN00440
C VARIABLE DICTIONARYMAN00450
C -----MAN00460
C CALLING ARGUMENTS:MAN00470
C -----MAN00480
C MAN00490
C MAN00500
C MAN00510
C MAN00520
C N NUMBER OF (X,Y) COORDINATES IN SAMPLE.MAN00530
C NOTE: MAXIMUM N=1000.MAN00540
C MAN00550

```

APPENDIX B: Fortran Coding For Mandel Subroutine

C	X	THE ONE-DIMENSIONAL ARRAY OF X OBSERVATIONS.	MAN00560
C			MAN00570
C	Y	THE ONE-DIMENSIONAL ARRAY OF Y OBSERVATIONS.	MAN00580
C			MAN00590
C	VXY	THE RATIO OF THE PRECISION VARIANCES IN MEASURED VALUES OF X AND Y. VXY=VAR(X)/VAR(Y).	MAN00600
C			MAN00610
C			MAN00620
C	RHO	CORRELATION OF X ERROR WITH Y ERROR. 0.0 <= RHO <= 1.0. RHO= 0.0 WHEN X AND Y ERRORS ARE INDEPENDENT.	MAN00630
C			MAN00640
C			MAN00650
C			MAN00660
C	PRINT	AN INDICATOR VARIABLE DETERMINING WHETHER ESTIMATED PARAMETERS ARE TO BE REPORTED AT USER'S TERMINAL. SETTING PRINT=0 SUPPRESSES OUTPUT, PRINT=1 DISPLAYS ESTIMATES. IN EITHER CASE, ALL PARAMETER ESTIMATES ARE AVAILABLE IN COMMON SPACE.	MAN00670
C			MAN00680
C			MAN00690
C			MAN00700
C			MAN00710
C			MAN00720
C			MAN00730
C			MAN00740
C			MAN00750
C			MAN00760
C			MAN00770
C			MAN00780
C			MAN00790
C	ALPHAT	('ALPHA HAT') ESTIMATED INTERCEPT OF REGRESSION LINE.	MAN00800
C			MAN00810
C			MAN00820
C	BETHAT	('BETA HAT') ESTIMATED SLOPE OF REGRESSION LINE.	MAN00830
C			MAN00840
C	STDALP	('STD. DEV. SUB ALPHA HAT') STANDARD ERROR OF INTERCEPT ESTIMATE.	MAN00850
C			MAN00860
C			MAN00870
C	SIDBET	('STD. DEV. SUB BETA HAT') STANDARD ERROR OF SLOPE ESTIMATE.	MAN00880
C			MAN00890
C			MAN00900
C	SIDDLI	('SID. DEV. SUB DELTA') ESTIMATED PRECISION OF X MEASUREMENTS.	MAN00910
C			MAN00920
C			MAN00930
C	STDEPS	('STD. DEV. SUB EPSILON') ESTIMATED PRECISION OF Y MEASUREMENTS. THE RESIDUAL ERROR IN CONVENTIONAL ORDINARY LEAST SQUARES PROCEDURE.	MAN00940
C			MAN00950
C			MAN00960
C			MAN00970
C	B	SLOPE ESTIMATE.	MAN00980
C			MAN00990
C	K	TANGENT OF ANGLE ALONG WHICH RESIDUALS MINIMIZED.	MAN01000
C			MAN01010
C	BK1	MANDEL'S STATISTIC GAUGING NECESSITY FOR SUBROUTINE MANDEL IN PLACE OF ORDINARY LEAST SQUARES PROCEDURE.	MAN01020
C			MAN01030
C			MAN01040
C	XO	A FUTURE MEASURED X VALUE.	MAN01050
C			MAN01060
C	YO	THE PREDICTED Y VALUE GIVEN XO.	MAN01070
C			MAN01080
C			MAN01090
C	STDVYO	STANDARD ERROR OF PREDICTED YO.	MAN01100
C			MAN01100

INTERNAL VARIABLES AVAILABLE WITHIN SUBROUTINE.

Y=F(X)

C	STDE	('STDV. DEV. SUB E') STANDARD ERROR OF ESTIMATE ON	MANO1110
C		TRANSFORMED SCALE.	MANO1120
C			MANO1130
C	SSUU	SUMS OF SQUARES, TRANSFORMED VARIABLE U.	MANO1140
C			MANO1150
C	SSVV	SUMS OF SQUARES, TRANSFORMED VARIABLE V.	MANO1160
C			MANO1170
C	C1-C5	CONVENIENT INTERMEDIATE VARIABLES USED TO CALCULATE	MANO1180
C		STANDARD ERROR OF ESTIMATE FOR FUTURE VALUES.	MANO1190
C			MANO1200
C		X=F(Y)	MANO1210
C		-----	MANO1220
C			MANO1230
C	ALPH2	('ALPHA HAT 2') ESTIMATED INTERCEPT OF REGRESSION	MANO1240
C		LINE.	MANO1250
C			MANO1260
C	BETH2	('BETA HAT 2') ESTIMATED SLOPE OF REGRESSION LINE.	MANO1270
C			MANO1280
C	SIDAL2	('SID. DEV. SUB ALPHA 2') STANDARD ERROR OF	MANO1290
C		INTERCEPT ESTIMATE.	MANO1300
C			MANO1310
C	STDB2	('STD. DEV. SUB BETA 2') STANDARD ERROR OF	MANO1320
C		SLOPE ESTIMATE.	MANO1330
C			MANO1340
C	STDD2	('STD. DEV. SUB DELTA 2') ESTIMATED PRECISION OF	MANO1350
C		THE Y MEASUREMENTS.	MANO1360
C			MANO1370
C	STDEP2	('STD. DEV. SUB EPSILON 2') ESTIMATED PRECISION OF	MANO1380
C		THE X MEASUREMENTS.	MANO1390
C			MANO1400
C	B2	SLOPE ESTIMATE.	MANO1410
C			MANO1420
C	K2	TANGENT OF ANGLE ALONG WHICH RESIDUALS MINIMIZED.	MANO1430
C			MANO1440
C	BK2	MANDEL'S STATISTIC GAUGING NECESSITY FOR SUBROUTINE	MANO1450
C		MANDEL IN PLACE OF ORDINARY LEAST SQUARES PROCEDURE.	MANO1460
C			MANO1470
C	YO	A FUTURE MEASURED Y VALUE.	MANO1480
C			MANO1490
C	XO	THE PREDICIED X VALUE GIVLN YO.	MANO1500
C			MANO1510
C	STDVXO	STANDARD ERROR OF PREDICTED XO.	MANO1520
C			MANO1530
C	SIDE2	('STD. DEV. SUB E 2') STANDARD ERROR OF ESTIMATE ON	MANO1540
C		TRANSFORMED SCALE.	MANO1550
C			MANO1560
C	SSUU2	SUMS OF SQUARES 2, TRANSFORMED VARIABLE U.	MANO1570
C			MANO1580
C	SSVV2	SUMS OF SQUARES 2, TRANSFORMED VARIABLE V.	MANO1590
C			MANO1600
C	C6-C10	CONVENIENT INTERMEDIATE VARIABLES USED TO CALCULATE	MANO1610
C		STANDARD ERROR OF ESTIMATE FOR FUTURE VALUES.	MANO1620
C			MANO1630
C			MANO1640
C		BOTH Y=F(X) AND X=F(Y)	MANO1650

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C          ----- MANO1660
C          SSXX  SUMS OF SQUARES, X. MANO1670
C          MANO1680
C          SSYY  SUMS OF SQUARES, Y. MANO1690
C          MANO1700
C          SSXY  CROSS PRODUCT SUMS OF SQUARES, X & Y. MANO1710
C          MANO1720
C          MANO1730
C          MANO1740
C----- MANO1750
C          AUXILLARY SUBROUTINES AND REQUIRED ARGUMENTS FOLLOW. MANO1760
C          MANO1770
C          SUBROUTINE REPORT(N,X,Y,RHO) MANO1780
C          MANO1790
C          WHERE CALLING ARGUMENTS ARE SAME AS ABOVE. THIS MANO1800
C          SUBROUTINE CONTAINS FORMAT STATEMENTS NECESSARY MANO1810
C          TO PRINT OUT PARAMETER ESTIMATES AND MAKES USE OF MANO1820
C          COMMON SPACE. MANO1830
C          MANO1840
C          SUBROUTINE CONF1(N) MANO1850
C          MANO1860
C          WHERE N IS THE SAMPLE SIZE. THIS SUBROUTINE MANO1870
C          ESTIMATES AND REPORTS THE CONFIDENCE INTERVALS FOR MANO1880
C          THE SLOPE, INTERCEPT, AND RESIDUAL ERROR PARAMETERS AND MANO1890
C          MAKES USE OF COMMON SPACE. MANO1900
C          MANO1910
C          SUBROUTINE CONF2(N,INPUT) MANO1920
C          MANO1930
C          WHERE N IS THE SAMPLE SIZE AND INPUT IS THE MENU MANO1940
C          SELECTION. THIS SUBROUTINE ESTIMATES AND REPORTS MANO1950
C          CONFIDENCE INTERVALS FOR Y AT SELECTED VALUES OF X MANO1960
C          AND VICE VERSA. IT ALSO MAKES USE OF COMMON SPACE. MANO1970
C          MANO1980
C          SUBROUTINE TTABLE(NDF,I,AREA,MODE) MANO1990
C          MANO2000
C          WHERE TTABLE REFERS TO STUDENT T DISTRIBUTION AND MANO2010
C          NDF = DEGREES OF FREEDOM, MANO2020
C          T = STUDENT T STATISTIC, MANO2030
C          AREA = AREA MEASURED FROM LEFT SIDE OF DISTRIBUTION, MANO2040
C          MODE = 1 FOR T IN AND AREA OUT, AND MANO2050
C          MODE=2 FOR AREA IN AND T OUT. MANO2060
C          MANO2070
C          SUBROUTINE CSTABL(NDF,CHISQ,AREA,MODE) MANO2080
C          MANO2090
C          WHERE CSTABL REFERS TO CHI-SQUARE DISTRIBUTION AND MANO2100
C          CHISQ = CHI-SQUARE STATISTIC, MANO2110
C          AREA = AREA MEASURED FROM LEFT SIDE OF DISIRIBUTION, MANO2120
C          MODE = 1 FOR CHISQ IN AND AREA OUT, AND MANO2130
C          MODE = 2 FOR AREA IN AND CHISQ OUT. MANO2140
C          MANO2150
C          SUBROUTINE ZTABLE(Z,AREA,MODE) MANO2160
C          MANO2170
C          MANO2180
C          MANO2190
C          MANO2200

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C
C      WHERE ZTABLE REFERS TO THE NORMAL DISTRIBUTION AND
C      Z = THE Z-SCORE STATISTIC
C      AREA = AREA MEASURED FROM LEFT SIDE OF DISTRIBUTION,
C      MODE = 1 FOR Z-SCORE STATISTIC IN AND AREA OUT,
C      MODE = 2 FOR AREA IN AND Z-SCORE STATISTIC OUT.
C
C      SUBROUTINE STAT(Y,N,AVGY,SDY,YMIN,YMAX,YSKEW,YKURT)
C
C      WHERE THE FOLOWING DESCRIPTIVE STATISTICS ARE
C      GENERATED FOR THE N OBSERVATIONS OF THE ARRAY Y:
C      AVGY = AVERAGE
C      SDY = STANDARD DEVIATION
C      YMIN = MINIMUM VALUE
C      YMAX = MAXIMUM VALUE
C      YSKFW = SKEW STATISTIC
C      YKURT = KURTOSIS STATISTIC
C
C=====
C
C      SUBROUTINE MANDEL(N,X,Y,VXY,RHO,PRINT)
C      COMMON ALPHA2,ALPHAT,B,B2,BETH2,BETHAT,BK1,BK2,
C      *C1,C2,C3,C4,C5,C6,C7,C8,C9,C10,K,K2,QUAN1,
C      *QUAN2,QUAN3,QUAN4,QUAN5,QUAN6,SUMX,SUMY,SSUU,SSUU2,SSVV2,
C      *SSVV,SSXX,SSXY,SSYY,STDAL2,STDALP,STOB2,STOBET,STDD2,STDDL2,
C      *STDE,STDE2,STDEP2,STDEPS,STDVXO,STDVYO,THETA,THETA2,VARXY,
C      *VARYX,XO,XBAR,YO,YBAR
C      DOUBLE PRECISION ALPHA2,ALPHAT,B,B2,BETH2,BETHAT,
C      *C1,C2,C3,C4,C5,C6,C7,C8,C9,C10,K,K2,PRINT,QUAN1,
C      *QUAN2,QUAN3,QUAN4,QUAN5,QUAN6,SUMX,SUMY,SSUU,SSUU2,SSVV2,
C      *SSVV,SSXX,SSXY,SSYY,STDAL2,STDALP,STOB2,STOBET,STDD2,STDDL2,
C      *STDE,STDE2,STDEP2,STDEPS,STDVXO,STDVYO,THETA,THETA2,VARXY,
C      *VARYX,XO,XBAR,YO,YBAR
C      REAL BK1,BK2,U(1000),V(1000),X(N),Y(N)
1      CONTINUE
C      IF(VXY.LE.O.)GO TO 30
C      IF(N.LT.3.OR.N.GT.1000)GO TO 40
C      IF(RHO.LT.O..OR.RHO.GT.1.)GO TO 50
C
C      INITIALIZE AND DETERMINE GENERAL STATISTICS
C
C      SUMX=0.
C      SUMY=0.
C      DO 10 I=1,N
10     SUMX=SUMX+X(I)
C      SUMY=SUMY+Y(I)
C      XBAR=SUMX/FLOAT(N)
C      YBAR=SUMY/FLOAT(N)
C      VARYX=VXY
C      VARYX=1./VARYX
C      THETA=0.
C      IF(RHO.GT.O.)THETA=RHO*DSQRT(VARYX)
C      THETA2=0.
C      IF(RHO.GT.O.)THETA2=RHO*DSQRT(VARYX)

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MANO2210
MANO2220
MANO2230
MANO2240
MANO2250
MANO2260
MANO2270
MANO2280
MANO2290
MANO2300
MANO2310
MANO2320
MANO2330
MANO2340
MANO2350
MANO2360
MANO2370
MANO2380
MANO2390
MANO2400
MANO2410
MANO2420
MANO2430
MANO2440
MANO2450
MANO2460
MANO2470
MANO2480
MANO2490
MANO2500
MANO2510
MANO2520
MANO2530
MANO2540
MANO2550
MANO2560
MANO2570
MANO2580
MANO2590
MANO2600
MANO2610
MANO2620
MANO2630
MANO2640
MANO2650
MANO2660
MANO2670
MANO2680
MANO2690
MANO2700
MANO2710
MANO2720
MANO2730
MANO2740
MANO2750

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SSXX=0. MANO2760
SSYY=0. MANO2770
SSXY=0. MANO2780
DO 20 I=1,N MANO2790
SSXX=(X(I)-XBAR)**2+SSXX MANO2800
SSYY=(Y(I)-YBAR)**2+SSYY MANO2810
20 SSXY=(X(I)-XBAR)*(Y(I)-YBAR)+SSXY MANO2820
C MANO2830
C CALCULATE PARAMETER ESTIMATES FOR Y=F(X) RELATIONSHIP MANO2840
C MANO2850
  QUAN1=(VARYX*SSXX-SSYY)**2-4.*(VARYX*SSXY-SSYY*THETA)*(THETA+SSXX-
  *SSXY) MANO2860
  QUAN1=DSQRT(QUAN1) MANO2870
  QUAN2=SSYY-VARYX*SSXX MANO2880
  QUAN3=2.*(SSXY+VARYX-SSYY*THETA) MANO2890
  K=(QUAN2+QUAN1)/QUAN3 MANO2910
  B=(SSXY+K*SSYY)/(SSXX+K*SSXY) MANO2920
  SSUU=SSXX+2.*K*SSXY+K*K*SSYY MANO2930
  SSVV=B*B*SSXX-2.*B*SSXY+SSYY MANO2940
  STDE=DSQRT(SSVV/(FLOAT(N-2))) MANO2950
  STDDL=STDE/DSQRT(B**2+VARYX-2.*B*THETA) MANO2960
  STDEPS=STDDL*DSQRT(VARYX) MANO2970
  STDALP=DSQRT(1./FLOAT(N)+(XBAR*XBAR*(1.+K*B)**2)/SSUU)*STDE MANO2980
  STDBET=(DABS(1.+K*B)/DSQRT(SSUU))*STDE MANO2990
  BETHAT=B MANO3000
  ALPHAT=YBAR-BETHAT*XBAR MANO3010
  BK1=K*B MANO3020
C MANO3030
C CALCULATE PARAMETER ESTIMATES FOR X=F(Y) RELATIONSHIP MANO3040
C (NOTE SUFFIX '2' ATTACHED TO VARIABLE NAMES) MANO3050
C MANO3060
  QUAN4=(VARXY*SSYY-SSXX)**2-4.*(VARXY*SSXY-SSXX*THETA2)*(THETA2*
  *SSYY-SSXY) MANO3070
  QUAN4=DSQRT(QUAN4) MANO3080
  QUAN5=SSXX-VARXY*SSYY MANO3090
  QUAN6=2.*(SSXY+VARXY-SSXX*THETA2) MANO3100
  K2=(QUAN5+QUAN4)/QUAN6 MANO3110
  B2=(SSXY+K2*SSXX)/(SSYY+K2*SSXY) MANO3120
  BK2=K2*B2 MANO3130
  SSUU2=SSYY+2.*K2*SSXY+K2*K2*SSXX MANO3140
  SSVV2=B2*B2*SSYY-2.*B2*SSXY+SSXX MANO3150
  STDE2=DSQRT(SSVV2/(FLOAT(N-2))) MANO3160
  STDD2=STDE2/DSQRT(B2**2+VARXY-2.*B2*THETA2) MANO3170
  STDEP2=STDD2*DSQRT(VARXY) MANO3180
  STDAL2=DSQRT(1./FLOAT(N)+(YBAR*YBAR*(1.+K2*B2)**2)/SSUU2)*STDE2 MANO3190
  STDB2=(DABS(1.+K2*B2)/DSQRT(SSUU2))*STDE2 MANO3200
  BETH2=B2 MANO3210
  ALPH2=XBAR-BETH2*YBAR MANO3220
  YO=ALPHAT+BETHAT*XO MANO3230
  C1=YBAR-BETHAT*XBAR MANO3240
  C2=BETHAT MANO3250
  C3=(BETHAT*STDDL)**2+STDE*STDE/FLOAT(N) MANO3260
  C4=(STDE*(1.+K*B))**2/SSUU MANO3270
  C5=(XO-XBAR)**2 MANO3280
  STDVYO=DSQRT(C3+C4+C5) MANO3290
  MANO3300

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1
36
1


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&'PREDETERMINED RATIO VAR(X-ERROR)/VAR(Y-ERROR) = ',F8.3,////, MANO3860
&T11,'PREDETERMINED CORRELATION OF X-ERROR WITH Y-ERROR = ',F6.3) MANO3870
  CALL STAT(X,N,AVGX,SDX,XMIN,XMAX,XSKEW,XKURT) MANO3880
  CALL STAT(Y,N,AVGY,SDY,YMIN,YMAX,YSKEW,YKURT) MANO3890
  WRITE(6,200) N, XMIN, YMIN, XMAX, YMAX, AVGX, AVGY, SDX, SDY, MANO3900
  &XSKEW, YSKEW, XKURT, YKURT MANO3910
200  FORMAT(///,T44,I4,' DATA POINTS',/,T36,33('-'),/,T43,'X', MANO3920
&T62,'Y',/,T36,14('-'),T55,14('-'),/,T26,'MINIMUM',T36,F14.6, MANO3930
&T55,F14.6,/,T26,'MAXIMUM',T36,F14.6,T55,F14.6,/,T29,'MEAN', MANO3940
&T36,F14.6,T55,F14.6,/,T15,'STANDARD DEVIATION',T36,F14.6,T55, MANO3950
&F14.6,/,T29,'SKEW',T40,F6.2,T59,F6.2,/,T25,'KURTOSIS',T40,F6.2, MANO3960
&T59,F6.2) MANO3970
  WRITE(6,300) ALPHAT, ALPH2, BETHAT, BETH2, STDEPS, STDDLT, BK1, MANO3980
  *BK2, STDALP, STDAL2, STDBET, STDB2 MANO3990
300  FORMAT(///,T43,'PARAMETER ESTIMATES',/,T36,33('-'),/,T39,'Y = ', MANO4000
&'F(X)',T58,'X = F(Y)',/,T36,14('-'),T55,14('-'),/,T24, MANO4010
&'INTERCEPT',T36,F14.6,T55,F14.6,/,T28,'SLOPE',T36,F14.6,T55, MANO4020
&F14.6,/,T1X,'RESIDUAL ERROR, S(YX) AND S(XY)',T36,F14.6,T55,F14.6, MANO4030
&/,T12,'MANDEL'S BK STATISTIC',T40,F6.2,T59,F6.2,/,T6,'STANDARD', MANO4040
&'ERROR OF INTERCEPT',T36,F14.6,T55,F14.6,/,T10,'STANDARD ERROR', MANO4050
&'OF SLOPE',T36,F14.6,T55,F14.6) MANO4060
  TALP1=-DABS(ALPHAT)/STDALP MANO4070
  CALL ITABLE(N-2,TALP1,SIGAL1,1) MANO4080
  TALPH2=-DABS(ALPH2)/STDAL2 MANO4090
  CALL ITABLE(N-2,TALPH2,SIGAL2,1) MANO4100
  TBETH1=-DABS(BETHAT)/STDBET MANO4110
  CALL ITABLE(N-2,TBETH1,SIGBH1,1) MANO4120
  TBETH2=-DABS(BETH2)/STDB2 MANO4130
  CALL ITABLE(N-2,TBETH2,SIGBH2,1) MANO4140
  WRITE(6,400) SIGAL1, SIGAL2, SIGBH1, SIGBH2 MANO4150
400  FORMAT(///,T37,'APPROXIMATE SIGNIFICANCE LEVELS',/,T36,33('-'),/, MANO4160
&T17,'NULL HYPOTHESIS',T38,'Y = F(X)',T59,'X = F(Y)',/,T16,17('-'), MANO4170
&T36,12('-'),T57,12('-'),/,T16,'INTERCEPT IS ZERO',T40,F5.3,T61, MANO4180
&F5.3,/,T20,'SLOPE IS ZERO',T40,F5.3,T61,F5.3) MANO4190
1  WRITE(6,500) MANO4200
500  FORMAT(///,T2,'SELECT ONE OF THE FOLLOWING OPTIONS',/,T5,'1. ', MANO4210
&'INTERVAL ESTIMATES FOR INTERCEPT, SLOPE, AND RESIDUAL ERROR', MANO4220
&/,T5,'2. INTERVAL ESTIMATES FOR TRUE Y AT SELECTED VALUE OF X',/ MANO4230
&,T5,'3. INTERVAL ESTIMATES FOR TRUE X AT SELECTED VALUE OF Y',/ MANO4240
&T5,'4. TERMINATE THIS RUN') MANO4250
  READ(5,*) INPUT MANO4260
  IF(INPUT.EQ.1) CALL CONF1(N) MANO4270
  IF((INPUT.EQ.2).OR.(INPUT.EQ.3)) CALL CONF2(N,INPUT) MANO4280
  IF((INPUT.LE.0).OR.(INPUT.GE.4)) RETURN MANO4290
  GO TO 1 MANO4300
  END MANO4310
C  MANO4320
  SUBROUTINE CONF1(N) MANO4330
  COMMON ALPH2,ALPHAT,B,B2,BETH2,BETHAT,BK1,BK2, MANO4340
  *C1,C2,C3,C4,C5,C6,C7,C8,C9,C10,K,K2,QUAN1, MANO4350
  *QUAN2,QUAN3,QUAN4,QUAN5,QUAN6,SUMX,SUMY,SSUU,SSUU2,SSVV2, MANO4360
  *SSVV,SSXX,SSXY,SSYY,STDAL2,STDALP,STDB2,STDBET,STDD2,STDDLT, MANO4370
  *STDE,STDE2,STDEP2,STDEPS,STDVXO,STDVYO,THETA,THETA2,VARXY, MANO4380
  *VARYX,XO,XBAR,YO,YBAR MANO4390
  DOUBLE PRECISION ALPH2,ALPHAT,B,B2,BETH2,BETHAT, MANO4400

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* C1, C2, C3, C4, C5, C6, C7, C8, C9, C10, K, K2, PRINT, QUAN1, MANO4410
* QUAN2, QUAN3, QUAN4, QUAN5, QUAN6, SUMX, SUMY, SSUU, SSUU2, SSVV2, MANO4420
* SSVV, SSXX, SSXY, SSYY, STDAL2, STDALP, STDB2, STDBET, STDD2, STDDLT, MANO4430
* SDF, SDF2, SDFP2, SDFPS, SDVXO, SDVYO, HETA, HETA2, VARXY, MANO4440
* VARYX, XO, XBAR, YO, YBAR MANO4450
DOUBLE PRECISION AL1MIN, AL2MAX, AL2MIN, ALPMAX, B2MAX, B2MIN, MANO4460
* BE1MAX, BETMIN, SDMAX, SDMIN, SEPMIN MANO4470
REAL C(10), SIGNIF(10), T(10) MANO4480
DATA SIGNIF/.005, .01, .025, .05, .1, .9, .95, .975, .99, .995/ MANO4490
WRITE(6,100) MANO4500
100 FORMAT(///,T18,'APPROXIMATE INTERVAL ESTIMATES FOR INTERCEPT',/, MANO4510
&T16,49(' '),/,T17,'TAIL AREA',T35,'Y = F(X)',T54,'X = F(Y)', MANO4520
&/,T16,11(' '),T32,14(' '),T51,14(' ')) MANO4530
DO 1 I=1,5 MANO4540
CALL TTABLE(N-2,T(I),1.-SIGNIF(I),2) MANO4550
AL1MIN=ALPHAT-T(I)*STDALP MANO4560
AL2MIN=ALPH2-T(I)*STDAL2 MANO4570
1 WRITE(6,200) SIGNIF(I), AL1MIN, AL2MIN MANO4580
200 FORMAT(T16,'LOWER ',F5.3,T32,F14.6,T51,F14.6) MANO4590
DO 2 I=1,5 MANO4600
J=6-I MANO4610
ALPMAX=ALPHAT+T(J)*STDALP MANO4620
AL2MAX=ALPH2+T(J)*STDAL2 MANO4630
2 WRITE(6,300) SIGNIF(J), ALPMAX, AL2MAX MANO4640
300 FORMAT(T16,'UPPER ',F5.3,T32,F14.6,T51,F14.6) MANO4650
WRITE(6,400) MANO4660
400 FORMAT(///,T20,'APPROXIMATE INTERVAL ESTIMATES FOR SLOPE',/,T16, MANO4670
&49(' '),/,T17,'TAIL AREA',T35,'Y = F(X)',T54,'X = F(Y)',/,T16, MANO4680
&11(' '),T32,14(' '),T51,14(' ')) MANO4690
DO 3 I=1,5 MANO4700
BETMIN=BETHAT-T(I)*STDBET MANO4710
B2MIN=BETH2-T(I)*STDB2 MANO4720
3 WRITE(6,200) SIGNIF(I), BETMIN, B2MIN MANO4730
DO 4 I=1,5 MANO4740
J=6-I MANO4750
BETMAX=BETHAT+T(J)*STDBET MANO4760
B2MAX=BETH2+T(J)*STDB2 MANO4770
4 WRITE(6,300) SIGNIF(J), BETMAX, B2MAX MANO4780
WRITE(6,500) MANO4790
500 FORMAT(///,T16,'APPROXIMATE INTERVAL ESTIMATES FOR RESIDUAL ERROR',MANO4800
&/,T16,49(' '),/,T17,'TAIL AREA',T37,'S(YX)',T56,'S(XY)', MANO4810
&/,T16,11(' '),T32,14(' '),T51,14(' ')) MANO4820
DO 5 I=1,5 MANO4830
CALL CSTABL(N-2,CHISO,1.-SIGNIF(I),2) MANO4840
SEPMIN=STDEPS/SQRT(CHISO/FLOAT(N-2)) MANO4850
SDMIN=STDDL/SQRT(CHISO/FLOAT(N-2)) MANO4860
5 WRITE(6,200) SIGNIF(I), SEPMIN, SDMIN MANO4870
DO 6 I=1,5 MANO4880
J=6-I MANO4890
CALL CSTABL(N-2,CHISO,SIGNIF(J),2) MANO4900
SEMAX=STDEPS/SQRT(CHISO/FLOAT(N-2)) MANO4910
SDMAX=STDDL/SQRT(CHISO/FLOAT(N-2)) MANO4920
6 WRITE(6,300) SIGNIF(J), SEMAX, SDMAX MANO4930
RETURN MANO4940
END MANO4950

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C
SUBROUTINE CONF2(N,INPUT)
COMMON ALPH2,ALPHAT,B,B2,BETH2,BETHAT,BK1,BK2,
* C1,C2,C3,C4,C5,C6,C7,C8,C9,C10,K,K2,QUAN1,
*QUAN2,QUAN3,QUAN4,QUAN5,QUAN6,SUMX,SUMY,SSUU,SSUU2,SSVV2,
*SSVV,SSXX,SSXY,SSYY,STDAL2,SIDALP,STDB2,STDBET,STDD2,STDDL2,
*STDE,STDE2,STDEP2,STDEPS,STDVXO,STDVYO,THETA,THETA2,VARXY,
*VARYX,XO,XBAR,YO,YBAR
DOUBLE PRECISION ALPH2,ALPHAT,B,B2,BETH2,BETHAT,
* C1,C2,C3,C4,C5,C6,C7,C8,C9,C10,K,K2,PRINT,QUAN1,
*QUAN2,QUAN3,QUAN4,QUAN5,QUAN6,SUMX,SUMY,SSUU,SSUU2,SSVV2,
*SSVV,SSXX,SSXY,SSYY,STDAL2,SIDALP,STDB2,STDBET,STDD2,STDDL2,
*SIDL,SIDL2,SIDEP2,SIDEPS,SIDVXO,SIDVYO,THETA,THETA2,VARXY,
*VARYX,XO,XBAR,YO,YBAR
DOUBLE PRECISION XLIM(10),YLIM(10)
REAL BK1,BK2,SIGNIF(10),I(10)
DATA SIGNIF/.005,.01,.025,.05,.1,.9,.95,.975,.99,.995/
DO 10 I=1,10
10 CALL TTABLE(N-2,I(1),SIGNIF(I),2)
IF(INPUT.EQ.3)GO TO 40
WRITE(6,100)
100 FORMAT(/,1X,'ENTER VALUE OF X AT WHICH COMPUTATIONS ARE TO BE ',
&'MADE')
READ(5,*)XO
YO=ALPHAT+BETHAT*XO
C5=(XO-XBAR)*.2
STDVYO=DSQRT(C3+C4+C5)
WRITE(6,200)YO,STDVYO
200 FORMAT(/,T21,'POINT ESTIMATE OF TRUE Y = ',F14.6,///,T21,'S(Y ',
&'FOR OBSERVED X) = ',F14.6)
WRITE(6,300)
300 FORMAT(//,T27,'APPROXIMATE INTERVAL ESTIMATES',/,T26,32('-'),/,
&T27,'TAIL AREA',T42,'Y FOR OBSERVED X',/,T26,11('-'),T42,16('-'))
DO 20 I=1,5
YLIM(I)=YO+T(I)*STDVYO
WRITE(6,400)SIGNIF(I),YLIM(I)
400 FORMAT(T26,'LOWER ',F5.3,T44,F14.6)
J=6
20 CONTINUE
DO 30 I=6,10
J=J-1
YLIM(I)=YO+T(I)*STDVYO
WRITE(6,500)SIGNIF(J),YLIM(I)
500 FORMAT(T26,'UPPER ',F5.3,T44,F14.6)
30 CONTINUE
RETURN
40 CONTINUE
WRITE(6,600)
600 FORMAT(/,1X,'ENTER VALUE OF Y AT WHICH COMPUTATIONS ARE TO BE ',
&'MADE')
READ(5,*)YO
XO=ALPH2+BETH2*YO
C10=(YO-YBAR)*.2
STDVXO=DSQRT(C8+C9*C10)
WRITE(6,700)XO,STDVXO
MANO4960
MANO4970
MANO4980
MANO4990
MANO5000
MANO5010
MANO5020
MANO5030
MANO5040
MANO5050
MANO5060
MANO5070
MANO5080
MANO5090
MANO5100
MANO5110
MANO5120
MANO5130
MANO5140
MANO5150
MANO5160
MANO5170
MANO5180
MANO5190
MANO5200
MANO5210
MANO5220
MANO5230
MANO5240
MANO5250
MANO5260
MANO5270
MANO5280
MANO5290
MANO5300
MANO5310
MANO5320
MANO5330
MANO5340
MANO5350
MANO5360
MANO5370
MANO5380
MANO5390
MANO5400
MANO5410
MANO5420
MANO5430
MANO5440
MANO5450
MANO5460
MANO5470
MANO5480
MANO5490
MANO5500

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700  FORMAT(/,I21,'POINT ESTIMATE OF TRUE X = ',F14.6,///,I21,'S(X ', MAN05510
      &'FOR OBSERVED Y)      = ',F14.6) MAN05520
      WRITE(6,800) MAN05530
800  FORMAT(/,I27,'APPROXIMATE INTERVAL ESTIMATES',/,I26,32('-'),/, MAN05540
      &I27,'TAIL AREA',I42,'X FOR OBSERVED Y',/,I26,11('-'),I42,16('-')) MAN05550
      DO 50 I=1,5 MAN05560
      XLIM(I)=XO+T(I)*STDVXO MAN05570
      WRITE(6,400)SIGNIF(I),XLIM(I) MAN05580
50   CONTINUE MAN05590
      J=6 MAN05600
      DO GO I=6,10 MAN05610
      J=J-1 MAN05620
      XLIM(I)=XO+T(I)*STDVXO MAN05630
      WRITE(6,500)SIGNIF(J),XLIM(I) MAN05640
60   CONTINUE MAN05650
      RETURN MAN05660
      END MAN05670
C
      SUBROUTINE CSTABL(NDF,CHISQ,AREA,MODE) MAN05680
      IF(NDF.LT.1) GO TO 3 MAN05690
      IF(MODE.NE.1) GO TO 2 MAN05700
      IF(CHISQ.LT.0.) GO TO 4 MAN05710
      IF(NDF.GT.200) GO TO 1 MAN05720
      AREA=AC(NDF,CHISQ) MAN05730
      RETURN MAN05740
1     X=2./(9.+NDF) MAN05750
      Z=((CHISQ/NDF)+.3333333-1.+X)/SQRT(X) MAN05760
      CALL ZTABLE(Z,AREA,1) MAN05770
      RETURN MAN05780
2     IF(MODE.NE.2) GO TO 5 MAN05790
      IF((AREA.LT.0.0001).OR.(AREA.GT.0.9999)) GO TO 6 MAN05800
      X=2./(9.+NDF) MAN05810
      CALL ZTABLE(Z,AREA,2) MAN05820
      CHISQ=AMAX1(0.,NDF*(1.-X+Z*SQRT(X))+3) MAN05830
      IF(NDF.GT.200) RETURN MAN05840
      CHISQ=CA(NDF,AREA,CHISQ) MAN05850
      RETURN MAN05860
3     WRITE(6,100) MAN05870
100  FORMAT('O','DEGREES OF FREEDOM IS LESS THAN ONE IN SUBROUTINE ', MAN05880
      &'CSTABL') MAN05890
      STOP MAN05900
4     WRITE(6,200) MAN05910
200  FORMAT('O','CHI-SQUARE VALUE IS NOT SPECIFIED CORRECTLY IN ', MAN05920
      &'SUBROUTINE CSTABL') MAN05930
      STOP MAN05940
5     WRITE(6,300) MAN05950
300  FORMAT('O','MODE OF OPERATION IS NOT SPECIFIED CORRECTLY IN ', MAN05960
      &'SUBROUTINE CSTABL') MAN05970
      STOP MAN05980
6     WRITE(6,400) MAN05990
400  FORMAT('O','AREA IS OUTSIDE LIMITS OF 0.0001 - 0.9999 IN ', MAN06000
      &'SUBROUTINE CSTABL') MAN06010
      STOP MAN06020
      END MAN06030
C
      END MAN06040
      MAN06050

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	FUNCTION AC(NDF,CHISQ)	MAN06060
	DF=NDF	MAN06070
	CALL CDIR(CHISQ,DF,AC,D,IER)	MAN06080
	IF(IER.EQ.0) RETURN	MAN06090
	WRITE(6,100)	MAN06100
100	FORMAT('O','ERROR IN CHI-SQUARE AREA CALCULATION IN SUBROUTINE ',	MAN06110
	&'CSTABL')	MAN06120
	STOP	MAN06130
	END	MAN06140
C	FUNCTION CA(NDF,AREA,CSTART)	MAN06150
	A(C)=AC(NDF,C)	MAN06160
	CA=CSTART	MAN06170
	ATRIAL=A(CA)	MAN06180
	KOUNT=0	MAN06190
1	KOUNT=KOUNT+1	MAN06200
	IF(KOUNT.GT.100) GO TO 3	MAN06210
	ERROR=ATRIAL-AREA	MAN06220
	IF(ABS(ERROR).LE.0.00001) RETURN	MAN06230
	IF(CA.LL.0.) CA=1.E-6	MAN06240
	C1=CA	MAN06250
	A1=ATRIAL	MAN06260
	C2=C1	MAN06270
2	C2=1.001*C2	MAN06280
	IF(C2/C1.GT.1.01) GO TO 3	MAN06290
	A2=A(C2)	MAN06300
	IF(A1.EQ.A2) GO TO 2	MAN06310
	SLOPE=(A2-A1)/(C2-C1)	MAN06320
	CA=C2+(AREA-A2)/SLOPE	MAN06330
	ATRIAL=A(CA)	MAN06340
	GO TO 1	MAN06350
3	WRITE(6,100)	MAN06360
100	FORMAT('O','EXCESSIVE CALCULATION TIME REQUIRED IN SUBROUTINE ',	MAN06370
	&'CSTABL')	MAN06380
	STOP	MAN06390
	END	MAN06400
C	SUBROUTINE ZTABLE(Z,AREA,MODE)	MAN06410
	DATA A1/.33267/, B1/.4361836/, B2/-.1201676/, B3/.937298/,	MAN06420
	&C1/2.515517/, C2/.802853/, C3/.010328/, D1/1.432788/,	MAN06430
	&D2/.189269/, D3/.001308/, SQRT2P/2.506628/	MAN06440
	IF(MODE.NE.1) GO TO 2	MAN06450
	AREA=1.	MAN06460
	ZABS=ABS(Z)	MAN06470
	IF(ZABS.GT.5.) GO TO 1	MAN06480
	X=1./(1.4A1+ZABS)	MAN06490
	Y=EXP(-ZABS**2/2)/SQRT2P	MAN06500
	AREA=1.-Y*X*(B1+X*(B2+X*B3))	MAN06510
1	IF(Z.GE.0.) RETURN	MAN06520
	AREA=1.-AREA	MAN06530
	RETURN	MAN06540
2	IF(MODE.NE.2) GO TO 4	MAN06550
	IF((AREA.LT.0.).OR.(AREA.GT.1.)) GO TO 5	MAN06560
	Z=5.	MAN06570
	A=AREA	MAN06580
		MAN06590
		MAN06600

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      IF(A.GT.0.5) A=1.-A
      IF(A.LT.0.0000004) GO TO 3
      X=SQRT(ALOG(1./A**2))
      Z=X-(C1*X+(C2*X+C3))/(1.+X*(D1+X*(D2+X*D3)))
3     IF (AREA.GE.0.5) RETURN
      Z=-Z
      RETURN
4     WRITE(G,100)
100    FORMAT('O','MODE OF OPERATION IS NOT SPECIFIED CORRECTLY IN ',
      &'SUBROUTINE ZTABLE')
      STOP
5     WRITE(G,200)
200    FORMAT('O','AREA IS NOT SPECIFIED CORRECTLY IN SUBROUTINE ZTABLE')
      STOP
      END
C
C
C .....
C
C     SUBROUTINE CDTR
C
C     PURPOSE
C     COMPUTES P(X) = PROBABILITY THAT THE RANDOM VARIABLE U,
C     DISTRIBUTED ACCORDING TO THE CHI-SQUARE DISTRIBUTION WITH G
C     DEGREES OF FREEDOM, IS LESS THAN OR EQUAL TO X. F(G,X), THE
C     ORDINATE OF THE CHI-SQUARE DENSITY AT X, IS ALSO COMPUTED.
C
C     USAGE
C     CALL CDTR(X,G,P,D,IER)
C
C     DESCRIPTION OF PARAMETERS
C     X - INPUT SCALAR FOR WHICH P(X) IS COMPUTED.
C     G - NUMBER OF DEGREES OF FREEDOM OF THE CHI-SQUARE
C     DISTRIBUTION. G IS A CONTINUOUS PARAMETER.
C     P - OUTPUT PROBABILITY.
C     D - OUTPUT DENSITY.
C     IER - RESULTANT ERROR CODE WHERE
C     IER= 0 --- NO ERROR
C     IER=-1 --- AN INPUT PARAMETER IS INVALID. X IS LESS
C     THAN 0.0, OR G IS LESS THAN 0.5 OR GREATER
C     THAN 2*10**(+5). P AND D ARE SET TO -1.E75.
C     IER=+1 --- INVALID OUTPUT. P IS LESS THAN ZERO OR
C     GREATER THAN ONE, OR SERIES FOR T1 (SEE
C     MATHEMATICAL DESCRIPTION) HAS FAILED TO
C     CONVERGE. P IS SET TO 1.E75.
C
C     REMARKS
C     SEE MATHEMATICAL DESCRIPTION.
C
C     SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED
C     DLGAM
C     NDIR
C
C     METHOD
C     REFER TO R.E. BARGMANN AND S.P. GHOSH, STATISTICAL
C     DISTRIBUTION PROGRAMS FOR A COMPUTER LANGUAGE,

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MAN06610
MAN06620
MAN06630
MAN06640
MAN06650
MAN06660
MAN06670
MAN06680
MAN06690
MAN06700
MAN06710
MAN06720
MAN06730
MAN06740
MAN06750
MAN06760
MAN06770
MAN06780
MAN06790
MAN06800
MAN06810
MAN06820
MAN06830
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MAN06880
MAN06890
MAN06900
MAN06910
MAN06920
MAN06930
MAN06940
MAN06950
MAN06960
MAN06970
MAN06980
MAN06990
MAN07000
MAN07010
MAN07020
MAN07030
MAN07040
MAN07050
MAN07060
MAN07070
MAN07080
MAN07090
MAN07100
MAN07110
MAN07120
MAN07130
MAN07140
MAN07150

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C	IBM RESEARCH REPORT RC-1094, 1963.	MANO7160
C		MANO7170
C	MANO7180
C		MANO7190
C	SUBROUTINE CDTR(X,G,P,D,IER)	MANO7200
C	DOUBLE PRECISION XX,DLXX,X2,DLX2,GG,G2,DLT3,THETA,THP1,	MANO7210
C	IGLG2,DD,111,SER,CC,XI,FAC,FLOG,TERM,GTH,A2,A,B,C,DT2,DT3,THPT	MANO7220
C		MANO7230
C	TEST FOR VALID INPUT DATA	MANO7240
C		MANO7250
C	IF(G-(.5-1.E-5)) 590,10,10	MANO7260
C	10 IF(G-2.E+5) 20,20,590	MANO7270
C	20 IF(X) 590,30,30	MANO7280
C		MANO7290
C	TEST FOR X NEAR 0.0	MANO7300
C		MANO7310
C	30 IF(X-1.E-8) 40,40,80	MANO7320
C	40 P=0.0	MANO7330
C	IF(G-2.) 50,60,70	MANO7340
C	50 D=1.E75	MANO7350
C	GO TO 610	MANO7360
C	60 D=0.5	MANO7370
C	GO TO 610	MANO7380
C	70 D=0.0	MANO7390
C	GO TO 610	MANO7400
C		MANO7410
C	TEST FOR X GREATER THAN 1.E+6	MANO7420
C		MANO7430
C	80 IF(X-1.E+6) 100,100,90	MANO7440
C	90 D=0.0	MANO7450
C	P=1.0	MANO7460
C	GO TO 610	MANO7470
C		MANO7480
C	SET PROGRAM PARAMETERS	MANO7490
C		MANO7500
C	100 XX=DBLE(X)	MANO7510
C	DLXX=DLOG(XX)	MANO7520
C	X2=XX/2.DO	MANO7530
C	DLX2=DLOG(X2)	MANO7540
C	GG=DBLE(G)	MANO7550
C	G2=GG/2.DO	MANO7560
C		MANO7570
C	COMPUTE ORDINATE	MANO7580
C		MANO7590
C	CALL DLGAM(G2,GLG2,I0K)	MANO7600
C	DD=(G2-1.DO)*DLXX-X2-G2+.6931471805599453 -GLG2	MANO7610
C	IF(DD-1.68D02) 110,110,120	MANO7620
C	110 IF(DD+1.68D02) 130,130,140	MANO7630
C	120 D=1.E75	MANO7640
C	GO TO 150	MANO7650
C	130 D=0.0	MANO7660
C	GO TO 150	MANO7670
C	140 DD=DEXP(DD)	MANO7680
C	D=SINGL(DD)	MANO7690
C		MANO7700

C	TEST FOR G GREATER THAN 1000.0	MAN07710
C	TEST FOR X GREATER THAN 2000.0	MAN07720
C		MAN07730
	150 IF(G-1000.) 160,160,180	MAN07740
	160 IF(X-2000.) 190,190,170	MAN07750
	170 P=1.0	MAN07760
	GO TO 610	MAN07770
	180 A=DLOG(XX/GG)/3.DO	MAN07780
	A=DEXP(A)	MAN07790
	B=2.DO/(9.DO+GG)	MAN07800
	C=(A-1.DO+B)/DSQRT(B)	MAN07810
	SC=SNGL(C)	MAN07820
	CALL NDTR(SC,P,DUMMY)	MAN07830
	GO TO 490	MAN07840
C		MAN07850
C	COMPUTE THETA	MAN07860
C		MAN07870
	190 K= 1DINI(G2)	MAN07880
	THETA=G2-DFLOAT(K)	MAN07890
	IF(THETA-1.D-8) 200,200,210	MAN07900
	200 THETA=0.DO	MAN07910
	210 THP1=THETA+1.DO	MAN07920
C		MAN07930
C	SELECT METHOD OF COMPUTING T1	MAN07940
C		MAN07950
	IF(THETA) 230,230,220	MAN07960
	220 IF(XX-10.DO) 260,260,320	MAN07970
C		MAN07980
C	COMPUTE T1 FOR THETA EQUALS 0.0	MAN07990
C		MAN08000
	230 IF(X2-1.68D02) 250,240,240	MAN08010
	240 T1=1.0	MAN08020
	GO TO 400	MAN08030
	250 T11=1.DO-DEXP(-X2)	MAN08040
	T1=SNGL(T11)	MAN08050
	GO TO 400	MAN08060
C		MAN08070
C	COMPUTE T1 FOR THETA GREATER THAN 0.0 AND	MAN08080
C	X LESS THAN OR EQUAL TO 10.0	MAN08090
C		MAN08100
	260 SER=X2*(1.DO/THP1 -X2/(THP1+1.DO))	MAN08110
	J=1	MAN08120
	CC=DFLOAT(J)	MAN08130
	DO 270 IT1=3,30	MAN08140
	XI=DFLOAT(IT1)	MAN08150
	CALL DLGAM(XI,FAC,IOK)	MAN08160
	TLOG= XI+DLX2-FAC-DLOG(XI+THETA)	MAN08170
	TERM=DEXP(TLOG)	MAN08180
	TERM=DSIGN(TERM,CC)	MAN08190
	SER=SER+TERM	MAN08200
	CC= -CC	MAN08210
	IF(DABS(TERM)-1.D-9) 280,270,270	MAN08220
	270 CONTINUE	MAN08230
	GO TO 600	MAN08240
	280 IF(SIR) 600,600,290	MAN08250

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290 CALL DLGAM(THP1,GTH,IOK)
      TLOG=THETA*DLX2+DLOG(SER) -GTH
      IF(TLOG+1.68D02) 300,300,310
300 T1=0.0
      GO TO 400
310 T11=DEXP(TLOG)
      T1=SNGL(T11)
      GO TO 400
C
C      COMPUTE T1 FOR THETA GREATER THAN 0.0 AND
C      X GREATER THAN 10.0 AND LESS THAN 2000.0
C
320 A2=0.0
      DO 340 I=1,25
          XI=DILOAT(I)
          CALL DLGAM(THP1,GTH,IOK)
          T11=- (13.00*XX)/XI +THP1*DLOG(13.00*XX/XI) -GTH-DLOG(XI)
          IF(T11+1.68D02) 340,340,330
330 T11=DEXP(T11)
          A2=A2+T11
340 CONTINUE
          A=1.0128205+THETA/156.00-XX/312.00
          B=DABS(A)
          C=-X2+THP1*DLX2+DLOG(B) -GTH-3.951243718581427
          IF(C+1.68D02) 370,370,350
350 IF (A) 360,370,380
360 C=-DEXP(C)
          GO TO 390
370 C=0.0
          GO TO 390
380 C=DEXP(C)
390 C=A2+C
          T11=1.00-C
          T1=SNGL(T11)
C
C      SELECT PROPER EXPRESSION FOR P
C
400 IF(G-2.) 420,410,410
410 IF(G-4.) 450,460,460
C
C      COMPUTE P FOR G GREATER THAN ZERO AND LESS THAN 2.0
C
420 CALL DLGAM(THP1,GTH,IOK)
      DT2=THETA*DLXX-X2-THP1*.6931471805599453 -GTH
      IF(DT2+1.68D02) 430,430,440
430 P=T1
          GO TO 490
440 DT2=DEXP(DT2)
          T2=SNGL(DT2)
          P=T1+T2+T2
          GO TO 490
C
C      COMPUTE P FOR G GREATER THAN OR EQUAL TO 2.0
C      AND LESS THAN 4.0
C

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MAN08260
MAN08270
MAN08280
MAN08290
MAN08300
MAN08310
MAN08320
MAN08330
MAN08340
MAN08350
MAN08360
MAN08370
MAN08380
MAN08390
MAN08400
MAN08410
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MAN08670
MAN08680
MAN08690
MAN08700
MAN08710
MAN08720
MAN08730
MAN08740
MAN08750
MAN08760
MAN08770
MAN08780
MAN08790
MAN08800

450	P=11	MAN08810
	GO TO 490	MAN08820
C		MAN08830
C	COMPUTE P FOR G GREATER THAN OR EQUAL TO 4.0	MAN08840
C	AND LESS THAN OR EQUAL TO 1000.0	MAN08850
C		MAN08860
460	DT3=0.00	MAN08870
	DO 480 I3=2,K	MAN08880
	THPI=DFLOAT(I3)*THETA	MAN08890
	CALL DLGAM(THPI,GTH,IOK)	MAN08900
	DLT3=THPI*DLX2-DLXX-X2-GTH	MAN08910
	IF(DLT3+1.68002) 480,480,470	MAN08920
470	DT3=DT3+DEXP(DLT3)	MAN08930
480	CONTINUE	MAN08940
	T3=SNGL(DT3)	MAN08950
	P=11-T3-T3	MAN08960
C		MAN08970
C	SET ERROR INDICATOR	MAN08980
C		MAN08990
490	IF(P) 500,520,520	MAN09000
500	IF(ABS(P)-1.E-7) 510,510,600	MAN09010
510	P=0.0	MAN09020
	GO TO 610	MAN09030
520	IF(1.-P) 530,550,550	MAN09040
530	IF(ABS(1.-P)-1.E-7) 540,540,600	MAN09050
540	P=1.0	MAN09060
	GO TO 610	MAN09070
550	IF(P-1.E-8) 560,560,570	MAN09080
560	P=0.0	MAN09090
	GO TO 610	MAN09100
570	IF((1.0-P)-1.E-8) 580,580,610	MAN09110
580	P=1.0	MAN09120
	GO TO 610	MAN09130
590	IER=-1	MAN09140
	D=-1.E75	MAN09150
	P=-1.E75	MAN09160
	GO TO 620	MAN09170
600	IER=+1	MAN09180
	P=1.E75	MAN09190
	GO TO 620	MAN09200
610	IER=0	MAN09210
620	RETURN	MAN09220
	END	MAN09230
C		MAN09240
C	SUBROUTINE DLGAM	MAN09250
C		MAN09260
C	PURPOSE	MAN09270
C	COMPUTES THE DOUBLE PRECISION NATURAL LOGARITHM OF THE	MAN09280
C	GAMMA FUNCTION OF A GIVEN DOUBLE PRECISION ARGUMENT.	MAN09290
C		MAN09300
C	USAGE	MAN09310
C	CALL DLGAM(XX,DLNG,IER)	MAN09320
C		MAN09330
C	DESCRIPTION OF PARAMETERS	MAN09340
C	XX - THE DOUBLE PRECISION ARGUMENT FOR THE LOG GAMMA	MAN09350

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C          FUNCTION. MANO9360
C          DLNG - THE RESULTANT DOUBLE PRECISION LOG GAMMA FUNCTION MANO9370
C          VALUE. MANO9380
C          IER - RESULTANT ERROR CODE WHERE MANO9390
C          IER= 0----NO ERROR. MANO9400
C          IER=-1----XX IS WITHIN 10**(-9) OF BEING ZERO OR XX MANO9410
C          IS NEGATIVE. DLNG IS SET TO -1.0D75. MANO9420
C          IER=+1----XX IS GREATER THAN 10**70. DLNG IS SET TO MANO9430
C          +1.0D75. MANO9440
C          MANO9450
C          REMARKS MANO9460
C          NONE MANO9470
C          MANO9480
C          SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED MANO9490
C          NONE MANO9500
C          MANO9510
C          METHOD MANO9520
C          THE EULER-MCLAURIN EXPANSION TO THE SEVENTH DERIVATIVE TERM MANO9530
C          IS USED, AS GIVEN BY M. ABRAMOWITZ AND I.A. STEGUN, MANO9540
C          'HANDBOOK OF MATHEMATICAL FUNCTIONS', U. S. DEPARTMENT OF MANO9550
C          COMMERCE, NATIONAL BUREAU OF STANDARDS APPLIED MATHEMATICS MANO9560
C          SERIES, 1966, EQUATION 6.1.41. MANO9570
C          MANO9580
C          ..... MANO9590
C          SUBROUTINE DLGAM(XX,DLNG,IER) MANO9600
C          DOUBLE PRECISION XX,ZZ,TERM,RZ2,DLNG MANO9610
C          IER=0 MANO9620
C          ZZ=XX MANO9630
C          IF(XX-1.D+10) 2,2,1 MANO9640
C          1 IF(XX-1.D+70) 8,9,9 MANO9650
C          MANO9660
C          SLE IF XX IS NEAR ZERO OR NEGATIVE MANO9670
C          MANO9680
C          2 IF(XX-1.D-9) 3,3,4 MANO9690
C          3 IER=-1 MANO9700
C          DLNG=-1.D75 MANO9710
C          GO TO 10 MANO9720
C          MANO9730
C          XX GREATER THAN ZERO AND LESS THAN OR EQUAL TO 1.D+10 MANO9740
C          MANO9750
C          4 TERM=1.DO MANO9760
C          5 IF(ZZ-18.DO) 6,6,7 MANO9770
C          6 TERM=TERM+ZZ MANO9780
C          ZZ=ZZ+1.DO MANO9790
C          GO TO 5 MANO9800
C          7 RZ2=1.DO/ZZ**2 MANO9810
C          DLNG =(ZZ-0.5D0)*DLOG(ZZ)-ZZ +0.9189385332046727 -DLOG(TERM)+ MANO9820
C          1(1.DO/ZZ)*(.8333333333333333D-1 -(RZ2*(.2777777777777777D-2 +(RZ2* MANO9830
C          2(.7936507936507936D-3 -(RZ2*(.5952380952380952D-3)))))) MANO9840
C          GO TO 10 MANO9850
C          MANO9860
C          MANO9870
C          XX GREATER THAN 1.D+10 AND LESS THAN 1.D+70 MANO9880
C          MANO9890
C          8 DLNG=ZZ*(DLOG(ZZ)-1.DO) MANO9900
    
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C          GO TO 10                                MAN09910
C          XX GREATER THAN OR EQUAL TO 1.D+70     MAN09920
C          MAN09930
C          MAN09940
C          9 IFR=11                                MAN09950
C          DING=1.D75                              MAN09960
C          .10 RETURN                              MAN09970
C          END                                     MAN09980
C          MAN09990
C          .....                                MAN10000
C          SUBROUTINE NDTR                          MAN10010
C          MAN10020
C          PURPOSE                                 MAN10030
C          COMPUTES Y = P(X) = PROBABILITY THAT THE RANDOM VARIABLE U, MAN10040
C          DISTRIBUTED NORMALLY(0,1), IS LESS THAN OR EQUAL TO X.    MAN10050
C          F(X), THE ORDINATE OF THE NORMAL DENSITY AT X, IS ALSO    MAN10060
C          COMPUTED.                                               MAN10070
C          MAN10080
C          USAGE                                       MAN10090
C          CALL NDTR(X,P,D)                                MAN10100
C          MAN10110
C          DESCRIPTION OF PARAMETERS                    MAN10120
C          X--INPUT SCALAR FOR WHICH P(X) IS COMPUTED.    MAN10130
C          P--OUTPUT PROBABILITY.                       MAN10140
C          D--OUTPUT DENSITY.                            MAN10150
C          MAN10160
C          REMARKS                                     MAN10170
C          MAXIMUM ERROR IS 0.0000007.                  MAN10180
C          MAN10190
C          SUBROUTINES AND SUBPROGRAMS REQUIRED         MAN10200
C          NONE                                          MAN10210
C          MAN10220
C          METHOD                                       MAN10230
C          BASED ON APPROXIMATIONS IN C. HASTINGS, APPROXIMATIONS FOR MAN10240
C          DIGITAL COMPUTERS, PRINCETON UNIV. PRESS, PRINCETON, N.J., MAN10250
C          1955. SEE EQUATION 26.2.17, HANDBOOK OF MATHEMATICAL    MAN10260
C          FUNCTIONS, ABRAMOWITZ AND STEGUN, DOVER PUBLICATIONS, INC., MAN10270
C          NLW YORK.                                       MAN10280
C          MAN10290
C          .....                                MAN10300
C          SUBROUTINE NDTR(X,P,D)                      MAN10310
C          MAN10320
C          AX=ABS(X)                                    MAN10330
C          T=1.0/(1.0+.2316419*AX)                     MAN10340
C          D=0.3989423*EXP(-X*X/2.0)                   MAN10350
C          P = 1.0 - D*T*((1.330274*T - 1.821256)*T + 1.781478)*T - MAN10360
C          1 0.3565638)*T + 0.3193815)                MAN10370
C          IF(X)1,2,2                                    MAN10380
C          1 P=1.0-P                                     MAN10390
C          2 RETURN                                     MAN10400
C          END                                          MAN10410
C          SUBROUTINE TTABLE(NDF,T,AREA,MODE)          MAN10420
C          IF(NDF.LI.1) GO TO 6                         MAN10430
C          MAN10440
C          MAN10450

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      IF(MODE.NE.1) GO TO 3
      IF(NDF.GE.200) GO TO 1
      IF(ABS(T).GT.(70.-FLOAT(NDF)/4.)) GO TO 2
      AREA=AT(NDF,T)
      RETURN
1     IF(NDF.GE.1000) GO TO 2
      A90=A1(90,1)
      A100=A1(100,T)
      AREA=A100-900.+(A90-A100)*(.01-1./NDF)
      AREA=AMAX1(0.,AMIN1(1.,AREA))
      RETURN
2     AREA=AZ(T)
      RETURN
3     IF(MODE.NE.2) GO TO 7
      IF((AREA.LT.0.0001).OR.(AREA.GT.0.9999)) GO TO 8
      IF(NDF.GE.200) GO TO 4
      T=TA(NDF,AREA)
      RETURN
4     A=AMAX1(AREA,(1.-AREA))
      IF((A.LE.0.99).AND.(NDF.GE.1000)) GO TO 5
      IF((A.LE.0.999).AND.(NDF.GE.5000)) GO TO 5
      I90=IA(90,AREA)
      I100=IA(100,AREA)
      T=I100-900.+(I90-I100)*(.01-1./NDF)
      RETURN
5     T=ZA(AREA)
      RETURN
6     WRITE(6,100)
100    FORMAT('O','DEGREES OF FREEDOM IS LESS THAN ONE IN SUBROUTINE ',
      &'TTABLE')
      STOP
7     WRITE(6,200)
200    FORMAT('O','MODE OF OPERATION IS NOT SPECIFIED CORRECTLY IN ',
      &'SUBROUTINE TTABLE')
      STOP
8     WRITE(6,300)
300    FORMAT('O','AREA IS OUTSIDE LIMITS OF 0.0001 - 0.9999 IN ',
      &'SUBROUTINE TTABLE')
      STOP
      END
      FUNCTION AT(NDF,T)
      DATA PI/3.141593/
      X=ATAN(T/SQRT(FLOAT(NDF)))
      COSX=COS(X)
      IF((NDF/2)*2.EQ.NDF) GO TO 1
      K=1
      SUM=0.
      IF(NDF.EQ.1) GO TO 4
      SUM=COSX
      IF(NDF.EQ.3) GO TO 4
      GO TO 2
1     K=0
      SUM=1.
      IF(NDF.EQ.2) GO TO 4
2     COEF=1.

```

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MAN10460
MAN10470
MAN10480
MAN10490
MAN10500
MAN10510
MAN10520
MAN10530
MAN10540
MAN10550
MAN10560
MAN10570
MAN10580
MAN10590
MAN10600
MAN10610
MAN10620
MAN10630
MAN10640
MAN10650
MAN10660
MAN10670
MAN10680
MAN10690
MAN10700
MAN10710
MAN10720
MAN10730
MAN10740
MAN10750
MAN10760
MAN10770
MAN10780
MAN10790
MAN10800
MAN10810
MAN10820
MAN10830
MAN10840
MAN10850
MAN10860
MAN10870
MAN10880
MAN10890
MAN10900
MAN10910
MAN10920
MAN10930
MAN10940
MAN10950
MAN10960
MAN10970
MAN10980
MAN10990
MAN11000

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FILE: MANDEL FORTRAN A NLW JERSEY DEPARTMENT OF TREASURY - IIUB DATA CENTER

```

1 STOP=NDF/2-1
DO 3 I=1,ISLIP
  J=2+I+K
  COFF=COFF*(J-1)/J
  SUM=SUM+COFF*COSX**J
  AT=(1+(K*2./PI))*(X+SUM*SIN(X))+(1-K)+SUM*SIN(X))/2.
  RETURN
END
FUNCTION TA(NDF,AREA)
Z=ZA(AREA)
TA=Z*(Z+Z**3)/(4.*NDF)+(3.*Z+Z**5.*Z**3+5.*Z**5)/(96.*NDF**2)
ATRIAL=AT(NDF,TA)
KOUNT=0
1 KOUNT=KOUNT+1
  IF(KOUNT.GT.100) GO TO 3
  ERROR=ATRIAL-AREA
  IF(ABS(ERROR).LE.0.000005) RETURN
  IF(TA.EQ.0.) TA=1.E-6
  T1=TA
  A1=ATRIAL
  T2=11
  T2=1.-001*T2
  IF(T2/T1.GT.1.01) GO TO 3
  A2=AT(NDF,T2)
  IF(A1.EQ.A2) GO TO 2
  SLOPE=(A2-A1)/(T2-T1)
  TA=T2+(AREA-A2)/SLOPE
  ATRIAL=AT(NDF,TA)
  GO TO 1
2 WRITE(6,100)
3 FORMAT('O', 'EXCESSIVE CALCULATION TIME IN SUBROUTINE TTABLE')
STOP
END
FUNCTION AZ(Z)
DATA A1/.33267/, B1/.4361836/, B2/-.1201676/, B3/.937298/,
&SORT2P/2.506628/
AZ=1.
ZABS=ABS(Z)
IF(ZABS.GT.5.) GO TO 1
X=1./(1.+A1*ZABS)
Y=EXP(-ZABS**2/2)/SORT2P
AZ=1.-Y*X*(B1+X*(B2+X*B3))
1 IF(Z.GE.0.) RETURN
AZ=1.-AZ
RETURN
END
FUNCTION ZA(AREA)
DATA C1/2.515517/, C2/.802853/, C3/.010328/, D1/1.432788/,
&D2/.189269/, D3/.001308/
A=AMIN1(AREA,(1.-AREA))
X=SQRT(ALOG(1./A**2))
ZA=X*(C1+X*(C2+X*C3))/(1.+X*(D1+X*(D2+X*D3)))
1 IF(AREA.GE.0.5) RETURN
ZA=-ZA
RETURN

```

MAN11010
MAN11020
MAN11030
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MAN11050
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MAN11080
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MAN11100
MAN11110
MAN11120
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MAN11140
MAN11150
MAN11160
MAN11170
MAN11180
MAN11190
MAN11200
MAN11210
MAN11220
MAN11230
MAN11240
MAN11250
MAN11260
MAN11270
MAN11280
MAN11290
MAN11300
MAN11310
MAN11320
MAN11330
MAN11340
MAN11350
MAN11360
MAN11370
MAN11380
MAN11390
MAN11400
MAN11410
MAN11420
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MAN11440
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MAN11470
MAN11480
MAN11490
MAN11500
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MAN11540
MAN11550

	END	MAN11560
	SUBROUTINE STAT(A,N,AVG,STDV,AMIN,AMAX,SKEW,CURT)	MAN11570
	DIMENSION A(N)	MAN11580
	DOUBLE PRECISION A1, D2, D3, D4, D1, SDN, SUM	MAN11590
	IF(N.LT.1) GO TO 3	MAN11600
	A1=A(1)	MAN11610
	AMIN=A1	MAN11620
	AMAX=A1	MAN11630
	AVG=A1	MAN11640
	STDV=0.	MAN11650
	SKEW=0.	MAN11660
	CURT=0.	MAN11670
	IF(N.LT.2) GO TO 4	MAN11680
	SUM=0.	MAN11690
	DO 1 I=1,N	MAN11700
	AI=A(I)	MAN11710
	IF(AI.LT.AMIN) AMIN=AI	MAN11720
	IF(AI.GT.AMAX) AMAX=AI	MAN11730
1	SUM=SUM+AI	MAN11740
	AVG=SUM/N	MAN11750
	D2=0.	MAN11760
	D3=0.	MAN11770
	D4=0.	MAN11780
	DO 2 I=1,N	MAN11790
	D1=A(I)-AVG	MAN11800
	D2=D2+D1**2	MAN11810
	D3=D3+D1**3	MAN11820
2	D4=D4+D1**4	MAN11830
	IF(D2.EQ.0.) GO TO 5	MAN11840
	STDV=DSQRT(D2/(N-1))	MAN11850
	SDN=DSQRT(D2/N)	MAN11860
	SKEW=D3/(N*SDN**3)	MAN11870
	CURT=D4/(N*SDN**4)-3.	MAN11880
	RETURN	MAN11890
3	WRITE(6,100)	MAN11900
100	FORMAT('O','N IS LESS THAN ONE IN SUBROUTINE STAT')	MAN11910
	STOP	MAN11920
4	WRITE(6,200)	MAN11930
200	FORMAT('O','***WARNING***',/,1X,'N EQUALS ONE IN SUBROUTINE STAT',	MAN11940
	*,/,1X,'STANDARD DEVIATION, SKEW, AND KURTOSIS ARE INDETERMINATE')	MAN11950
	RETURN	MAN11960
5	WRITE(6,300)	MAN11970
300	FORMAT('O','***WARNING***',/,1X,'STANDARD DEVIATION IS ZERO IN ',	MAN11980
	*'SUBROUTINE STAT',/,1X,'SKEW AND KURTOSIS COEFFICIENTS ARE ',	MAN11990
	*'INDETERMINATE')	MAN12000
	RETURN	MAN12010
	END	MAN12020