

# A Flood-Frequency Relation Based on Regional Record Maxima

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GEOLOGICAL SURVEY PROFESSIONAL PAPER 434-F

*Prepared in cooperation with the  
U.S. Atomic Energy Commission*



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By P. H. CARRIGAN, JR.

STATISTICAL STUDIES IN HYDROLOGY

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## STATISTICAL STUDIES IN HYDROLOGY

# A FLOOD-FREQUENCY RELATION BASED ON REGIONAL RECORD MAXIMA

By P. H. CARRIGAN, JR.

### ABSTRACT

Flood intensity varies randomly in both time and space over a region. In flood-frequency analysis the regional variation is commonly related, through regression analysis, to variations in basin and climatic characteristics. Thus, the sample size is reduced to one time sample applicable to the entire region. This type of analysis virtually limits the recurrence-interval estimate for the maximum annual flood to the length of time for which records are available.

Techniques of analysis introduced in this paper increase the sample size by taking account of the random variations of flood intensity in both time and space. The magnitude of increase in the sample size is limited by the statistical dependence between annual flood records for the region; that is, the less the correlation between records the greater the effective sample size. As a result of increasing the sample size, the recurrence interval for the maximum event is increased; the increase may conceivably be as great as the number of station-years of record, if the records are independent.

The recurrence interval for the maximum annual flood in the region is estimated by a computer-simulation model. In order to apply the model, the assumption is made that the sets of concurrent records for a region are reducible to identical distributions by use of an appropriate scaling factor.

The dependence among records is preserved by including in the model the sample correlation coefficients for all pairs of record sets. By means of a principal-component transform, the matrix of correlation coefficients is used in connection with a normal random-number generator to generate synthetic sets of normally distributed flood intensities equal in number and duration to observed sets of records. The maximum event from each synthetic record is found, and the maxima are ranked in descending order. The exceedance probabilities (reciprocals of recurrence intervals) of the ordered maxima are repeatedly determined to estimate the average normal exceedance probability for each order. The scheme of computing exceedance probabilities of known distribution for ordered maxima selected from synthetic records of known distribution provides an estimate of probability that is independent of the statistical distribution of the observed floods.

The construction of a flood-frequency relation for any station in the region can be accomplished by graphically relating the observed ordered maxima, reduced to a common base, to the

corresponding estimated recurrence interval. The procedure of constructing a flood-frequency relation is illustrated by applying it to floods in the Big Lost River basin near the National Reactor Testing Station in southeastern Idaho.

### INTRODUCTION

Current practices of regional flood-frequency analysis define regions in which the records of annual hydrologic extremes have, for the most part, identical distributions. The method of analysis is that of multiple regression, whereby floods of specific recurrence intervals or the statistical parameters of flood distributions are related to basin and climatic characteristics. The method composites the experience over a region but, except by extension based on the assumed distribution of events, it produces results that are limited virtually to the length of the period of sampling. Thus, no advantage is taken of the facts that spatial sampling to some degree is equivalent to time sampling and that some of the experience within the region may furnish information about the expectancy of events for a period much longer than the actual period of record. Work by Conover and Benson (1963) suggests that this information can be realized by appropriate treatment of combined records from several streamflow stations in the defined area. In the Conover and Benson method of analysis, the process is assumed to be ergodic, so that spatial sampling is equivalent to time sampling of extreme events. The result is to increase the size of the sample to  $nk$  events in a region ( $k$ =years of record,  $n$ =number of sets of record), provided the records are independently and identically distributed.

If the Conover and Benson method of regional flood-frequency analysis were applied to records of annual extremes, the records could be reduced to identical distributions by normalizing them to the magnitudes

of the mean annual events at each streamflow station. However, the records for a region usually are not independent but are correlated. Hence, another method of analysis must be developed for treating the combined information from several identically distributed sets of records. This report presents a procedure to estimate exceedance probabilities of floods from records that are dependent and identically distributed. The maximum recurrence interval for the region, estimated by this procedure, lies between  $k+1$  and  $f(\rho)k+1$  where  $f(\rho)$  is a function of the correlation between record sets ( $1 \leq f(\rho) \leq n$ ) for the range in correlation coefficients ( $0 \leq \rho \leq 1$ ).

The scheme of Conover and Benson for treating the combined information from independent and identically distributed records is to compute expected exceedance probabilities (independent of the distribution) for the ordered record maxima. This scheme is the key to developing regional frequency relations for dependent records.

Once a region having identically distributed sets of concurrent annual flood records is defined, the regional flood-frequency relation is developed as follows:

1. The records are reduced to a common base.
2. The maximum events from each reduced record are ordered.
3. The correlation matrix for the records is computed.
4. Computer routines developed in this study are used to provide, through simulation algorithms that utilize the correlation matrix, estimates of exceedance probabilities associated with ordered maxima from  $n$ , number of  $k$ -year records.
5. The frequency relation is defined graphically by relating exceedance probabilities to the ordered maxima. Reversing the transformation process that reduces records to an identical distribution produces regionalized frequency relations for annual floods at individual stations.

After briefly illustrating the Conover and Benson method of obtaining exceedance probabilities for independent flood records, this report introduces the mathematical and computer-simulation foundations for estimating exceedance probabilities for dependent flood records. Then follow discussions on the application of the procedure to simple models which help to describe the influence of the number of records, correlation coefficients, and record length on exceedance probabilities; the development of regional flood-frequency relations in the Big Lost River basin, Idaho; and the extension of the procedure to incomplete and to serially correlated records. The complete computer program is presented at the end of the report.

## ACKNOWLEDGMENTS

This study of flood-frequency relations was part of a research project on the statistical occurrence of extreme hydrologic events in relation to the design and selection of sites for nuclear reactor facilities. The work was done on behalf of the Division of Reactor Development and Technology, U.S. Atomic Energy Commission.

Early considerations for estimating the probability of occurrence of rare events favored analytical approaches. However, after discussions with W. J. Conover of Kansas State University, it was decided to discard methods relying on analytical approaches. U.S. Geological Survey personnel made the following contributions to the study: M. A. Benson recommended use of Monte Carlo techniques; discussions with N. C. Matalas resulted in much of the structuring in the Monte Carlo model; E. J. Gilroy drew attention to the use of a moving-average scheme to simulate multivariate distributions; and M. S. Hellman kindly supplied an improved computer algorithm for finding the eigenvectors of a correlation matrix.

## CONOVER AND BENSON METHOD OF DETERMINING PROBABILITIES

Assume that  $n$  concurrent sets of  $k$  annual extremes are obtained from  $n$  identical and independent distributions. The maximum events, one from each record, are ordered from highest to lowest. The probability that another event  $f$  exceeds these ordered maxima

$$\{f_i; i=1, 2, \dots, n\} \text{ is}$$

$$P[f > f_i] = \sum_{m=0}^{i-1} \frac{n!}{(n-m)! k \prod_{j=0}^m \left(n + \frac{1}{k} - j\right)}$$

for  $i=1, 2, \dots, n$  (Conover and Benson, 1963, p. E159). *Example.*—The following three sets of four random events  $\{x_{ij}; i=1, 2, 3 \text{ and } j=1, 2, 3, 4\}$  are from independent and identical distributions:

		$x_{ij}$		
		$[n=3, k=4]$		
$j$	$i$			
	1	2	3	
1	3	69	3	
2	38	24	48	
3	17	61	60	
4	32	30	83	

The ordered record maxima  $\{f_i; i=1, 2, 3\}$  are, respectively, 83, 69, and 38. The probability that another event  $f$  exceeds the second highest maxima is

$$\begin{aligned}
 P[f > 69] &= \sum_{m=0}^{2-1} \frac{3!}{(3-m)!k \prod_{j=0}^m \left(3 + \frac{1}{4} - j\right)} \\
 &= \frac{3!}{3!4 \left(3 + \frac{1}{4}\right)} + \frac{3!}{2!4 \left(3 + \frac{1}{4}\right) \left(3 + \frac{1}{4} - 1\right)} \\
 &= \frac{1}{13} + \frac{3}{29.25} = 0.180.
 \end{aligned}$$

**METHOD FOR DETERMINING PROBABILITIES FROM DEPENDENT RECORDS**

The probability of another random event  $f$  exceeding one of the ordered record maxima  $f_i$  where the records are from dependent distributions cannot be analytically determined. The probability can be determined, however, through use of Monte Carlo simulation techniques. In this application of simulation for dependent records, it is assumed that the probability of another random event exceeding an ordered record maximum is independent of the identical distributions for which the records of random events are a sample. Such an assumption has been verified by Conover and Benson (1963) and proven by Conover (1965) for independent random events.

The simulation algorithm, which incorporates a normal random-number component having zero mean and unit variance, generates  $n$  sets of  $k$  events that simulate the dependence among  $n$  records of  $k$  hydrologic extremes. The maximum event  $y_i$  from each set of normally distributed variables  $\{x_{i1}, x_{i2}, \dots, x_{ik}; i=1, 2, \dots, n\}$  is found, and the maxima are ordered from highest to lowest, so that  $y_1 > y_2 > \dots > y_n$ . All  $y_i$  are from normal distributions having zero mean and unit variance, so their exceedance probabilities are given by

$$P[y > y_i] = \frac{1}{\sqrt{2\pi}} \int_{y_i}^{\infty} e^{-\frac{y^2}{2}} dy, \text{ for } i=1, 2, \dots, n.$$

By repeated generation of this simulated set of records, estimates of the average exceedance probabilities for the ordered maxima are obtained.

**SIMULATION MODEL**

The simulated sets of records are generated by the normal multivariate model

$$X = B\epsilon,$$

where  $X$  = matrix of  $n$  groups of  $k$  events,

$\epsilon = n \times k$  matrix of independent normal random numbers with zero mean and unit variance, and

$B = E\lambda = n \times n$  principal-component transform matrix for the records of annual extremes, in which

$E$  = eigenvector matrix associated with the correlation matrix  $R$  obtained from  $n$  records of  $k$  hydrologic events, and

$\lambda$  = diagonal matrix whose  $n$  diagonal elements are the square roots of the eigenvalues for  $E$ .

(See Morrison, 1967, p. 221-247.) This model is based on techniques of synthetic hydrology developed by Matalas (1967, p. 940).

After generation of  $X$ , the  $n$  maxima are selected and ordered and the exceedance probabilities are computed. The algorithm for computing normal exceedance probabilities is that of Zelen and Severs (1965, p. 932, eq. 26.2.19).

**FLOW CHART FOR SIMULATION**

The average exceedance probability

$$\begin{aligned}
 p_i &= P[y > y_i; y_1 > y_2 > \dots > y_n] \\
 &= \frac{1}{1,000} \sum_{m=1}^{1,000} \frac{1}{\sqrt{2\pi}} \int_{y_{i,m}}^{\infty} e^{-\frac{y^2}{2}} dy
 \end{aligned}$$

was estimated through use of a digital computer. The main program flow chart is shown in figure 1. Routines for computer input-output listings of matrices, matrix operations, and generation of random numbers are adaptations of subroutines developed by IBM (1968) for use in their 360-series computer systems. Formal declarations of most subroutines have been eliminated from the program to increase efficiency. A complete listing of the Fortran language program for the computations is included in the computer program at the end of this report. The parameters  $n$  and  $k$  may each be as large as 100 using the present configuration of the IBM 360/65 computing system (380,000 bytes available storage) operated by the U.S. Geological Survey.

**INFLUENCE OF INTERRECORD CORRELATION ON EXCEEDANCE PROBABILITY**

An  $n$ -record correlation matrix contains  $n(n-1)/2$  correlation coefficients. If  $n > 3$ , problems in management of matrix elements begin to obscure the study of the influence of interrecord correlation on exceedance probabilities. Hence, most simulations of exceedance probabilities were derived from the simple, yet informative, 3-record correlation matrices.



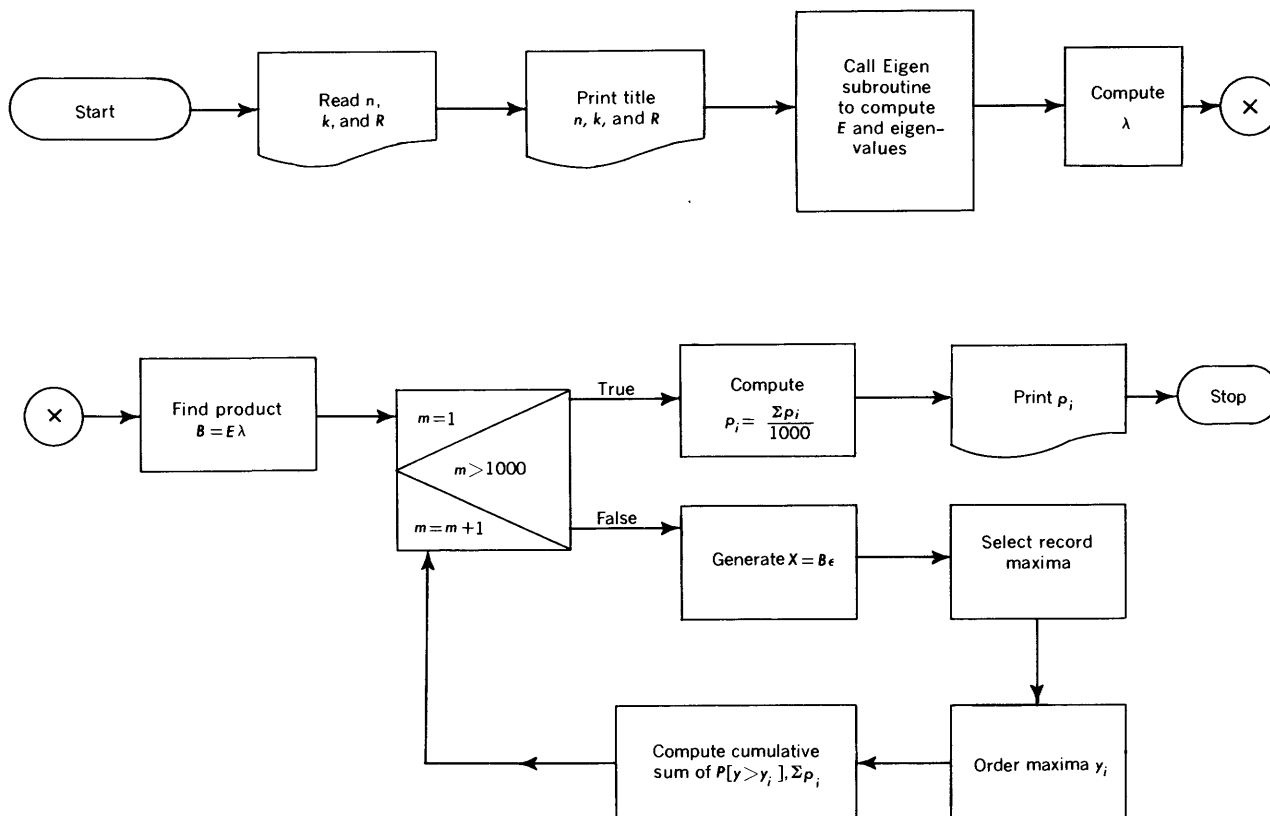


FIGURE 1.—Flow chart for program to simulate  $\{p_i(k); i=1, 2, \dots n\}$ .

**EQUICORRELATION MATRIX**

Using the equicorrelation matrix  $R$  (all correlation coefficients equal)

$$\begin{bmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{bmatrix}$$

for  $0 \leq \rho \leq 1$ , the expected probabilities  $\{p_i; i=1, 2, 3\}$  were simulated for  $k=10, 20, \dots 50$ .

As shown in figure 2,  $p_i$  for any order  $i$  changes monotonically, as had been anticipated—very gradually from the theoretical probability at  $\rho=0$  (three independent records) to about  $\rho=0.8$ , and then rapidly to the common theoretical probability at  $\rho=1$  (three identical records). Comparison of the simulated probability to the theoretical probability at  $\rho=0$  and at  $\rho=1$ , shown in figure 2, indicates that 1,000 iterations is sufficient for estimates of  $p_i$ . Similar monotonic variations in  $\{p_i; i=1, \dots n\}$  between  $\rho=0$  and  $\rho=1$  occurred for  $n=2$  and  $n=5$  where  $k=10, 20, \dots 50$ .

**VERIFICATION THAT  $p_i$  IS DISTRIBUTION FREE**

The probabilities  $p_i$  for  $n=2, k=10$ , and  $\rho=0, 0.1, 0.2, \dots 1.0$  were simulated by the previously de-

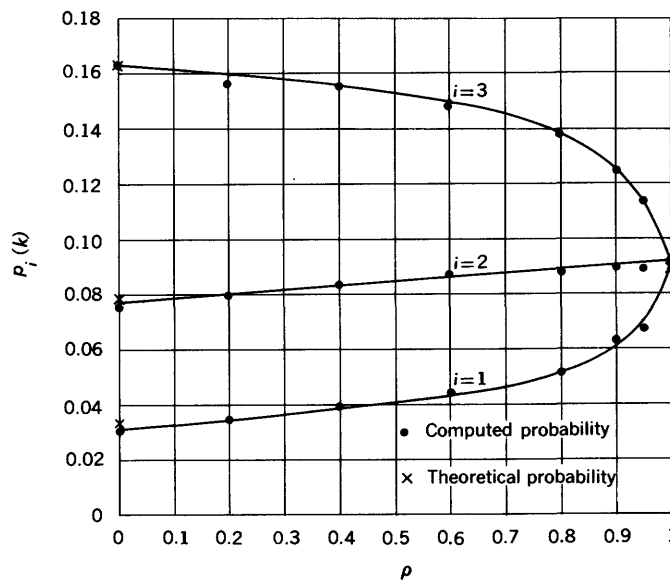


FIGURE 2.—Variation of  $\{p_i(k); i=1, 2, 3\}$  with  $\rho$  for  $k=10$ .

scribed normal variate generating model. The data for these simulation runs are plotted with their related curves in figure 3.

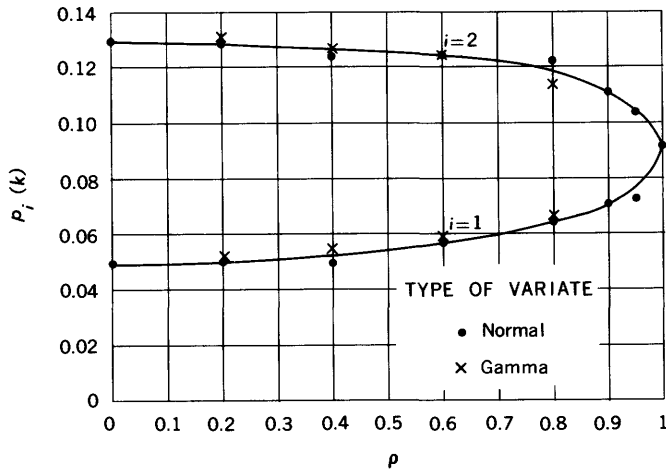


FIGURE 3.—Variation of  $p_1(k)$  and  $p_2(k)$  with  $\rho$  for bivariate gamma and normal distributions of annual extremes;  $k=10$ .

Another model for simulating a multivariate distribution  $\{z_{i,j}; i=1, 2, \dots, n \text{ and } j=1, 2, \dots, k\}$  is to let

$$z_{i,j} = \sum_{l=1}^m u_{(i-1)d+l,j}$$

in which  $u_{(i-1)d+l,j}$  = exponentially distributed independent random numbers having unit mean and unit variance, and

$z_{i,j}$  = random numbers following a marginal gamma distribution having mean  $m$  and variance  $m$ , for which the correlation coefficient is

$$\rho(z_i, z_{i+1}) = \frac{m-d}{m}$$

Using this simulation model the mean exceedance probabilities  $\{p_i; i=1, 2\}$  for the ordered maxima from  $z_{i,j}$  (for  $m=5, n=2$ , and  $k=10$ ) were estimated for  $\rho=0.2, 0.4, 0.6$ , and  $0.8$  ( $d=4, 3, 2, 1$ ); the coefficient of skew for  $z_{i,j}$  was  $0.89$  (the mean coefficient for a 50-record sample of annual floods was  $0.9$ ).

Results for this bivariate gamma method of estimating  $p_i$  are shown in figure 3. Within the experimental error experienced, the estimates appear to confirm the assumption that  $p_i$  is distribution free.

VARIATION OF  $p_i$  WITH  $k$  AND  $n$

The probability  $p_i$  varies smoothly and monotonically with either  $k$  or  $n$ . (See figs. 4 and 5 for examples.) The variation with  $k$  of  $p_i$  for  $0 < \rho < 1$  and  $n$  constant is similar to the variation of the theoretical probability ( $\rho=0$  and  $1$ ) with  $k$ . The relation between  $p_i$  and  $n$  for constant  $k$  shifts from being highly curvilinear for  $\rho$  near zero to being nearly horizontal for  $\rho$  near 1.

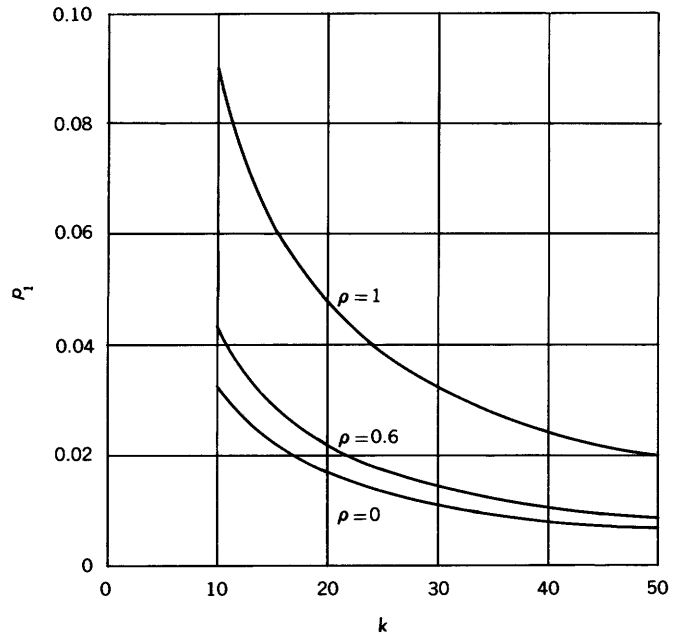


FIGURE 4.—Variation of  $p_1$  with  $k$  for  $n=3$  and  $\rho=0, 0.6$ , and  $1$ .

GENERAL CORRELATION MATRIX

The relative insensitivity of  $p_i$  to changes in  $\rho$  (fig. 2) suggests the use of a simple parameter

$$\bar{\rho}_n = \frac{2}{n(n-1)} \sum_{i=1}^{n-1} \sum_{j=i+1}^n \rho_{ij}$$

This parameter weights the central tendency of correlation coefficients in the general correlation matrix ( $\rho_{ij}$  are not all equal) and replaces  $\rho$  in estimating  $p_i'$  (the prime referring to a general matrix) from relations such as those in figure 2. Results of tests showed that all the variation in  $p_i'$  could not be accounted for by variations in  $\bar{\rho}_n$ . A parameter that described the dispersion of the coefficients in  $R$  about  $\bar{\rho}_n$ , such as a mean deviation

$$\alpha_n = \frac{2}{n(n-1)} \sum_{i=1}^{n-1} \sum_{j=i+1}^n |\rho_{ij} - \bar{\rho}_n|$$

or a mean square deviation

$$\delta_n^2 = \frac{2}{n(n-1)} \sum_{i=1}^{n-1} \sum_{j=i+1}^n (\rho_{ij} - \bar{\rho}_n)^2,$$

seemed necessary to explain variations in  $p_i'$  more fully.

Experiments using third-order correlation matrices were performed to define the variation of  $\{p_i'; i=1, 2, 3\}$  with increases in  $\alpha_3$  or  $\delta_3^2$  for constant  $\bar{\rho}_3$  (see fig. 6). Under these conditions,  $p_i'$  was found to increase

monotonically for  $i=1$  and  $i=2$  and to decrease monotonically for  $i=3$ .

Under the constraints that  $R$  be positive, semidefinite (see Morrison, 1967, p. 60),  $\bar{\rho}_n$  be constant, and  $0 \leq \rho_{ij} \leq 1$ , a maximum  $\alpha_n$  exists. For  $\bar{\rho}_3 \geq \frac{1}{3}$ ,  $\alpha_3$  is maximized if one of the  $\rho_{ij}=1$ , say  $\rho_{12}=1$ , and the other coefficients

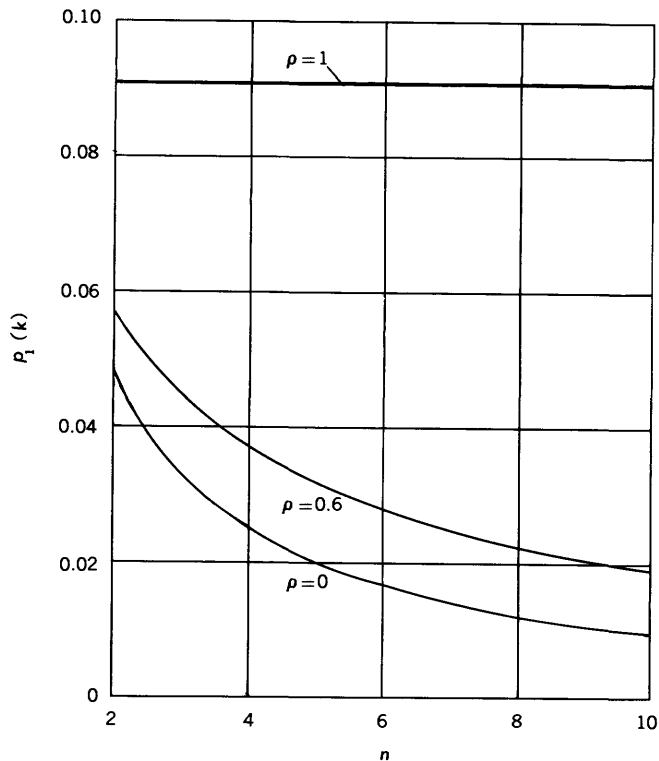


FIGURE 5.—Variation of  $p_1(k)$  with  $n$  for  $k=10$  and  $\rho=0, 0.6$ , and  $1$ .

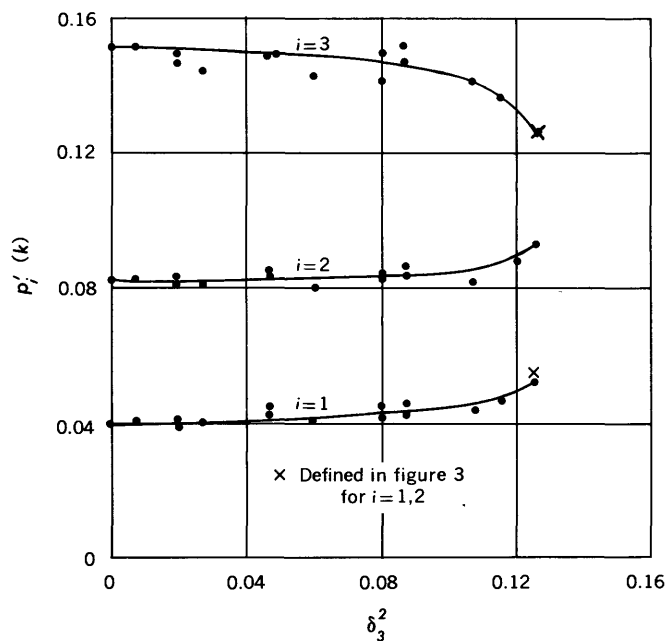


FIGURE 6.—Effect of mean square deviation  $\delta_3^2$  on  $\{p_i'(k); i=1, 2, 3\}$  for  $k=10$  and  $\rho=0.5$ .

are as follows:

$$\rho_{13} = \rho_{23} = \frac{3\bar{\rho}_3 - 1}{2};$$

hence,  $\rho_{13} = \rho_{23} = \rho$  for  $n=2$ . For  $\bar{\rho}_3 < \frac{1}{3}$ ,  $\alpha_3$  is maximized if  $\rho_{12} = 3\bar{\rho}_3$  and  $\rho_{13} = \rho_{23} = 0$ . The maximum is

$$\alpha_{3 \max} = 2(\rho_{12} - \bar{\rho}_3).$$

These conditions for which  $\alpha_3$  is a maximum obtain the maximum for  $\delta_3^2$ . The upper limit for  $p_i'$  at a maximum  $\delta_3^2$  is shown in figure 7.

The variation of  $p_i'$  with  $\delta_3^2$ , shown in figure 6, brings out a point about the maxima. For  $i=1$ ,  $p_1'$  increases to the limit  $p_1$  for  $n=2$ ; for  $i=3$ ,  $p_3'$  decreases to the limit  $p_2$  for  $n=2$ ; but  $p_2'$  does not reach either limit. If  $\rho_{12}=1$ , then only two ordered maxima occur, one in records 1 and 2 and one in record 3. One of the maxima may be first order and the other may be third order; the second-order maximum does not really exist. It would appear in figure 6 that  $p_2'$  is approaching the limit for  $\rho=1$  ( $p_i=0.0909$  for  $k=10$ ). Results from two other sets of tests—for  $k=50$  and  $n=3$  and for  $k=10$  and  $n=5$ —indicate that the variations of  $p_i'$  with  $\bar{\rho}_n$  and  $\delta_3^2$  are much like those shown in figures 6 and 7.

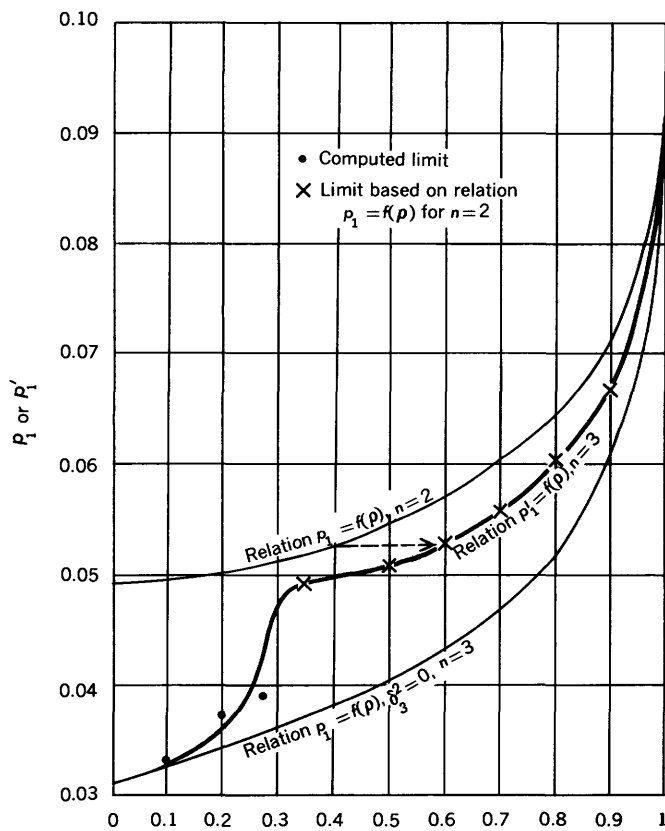


FIGURE 7.—Upper limit of  $p_i'$  for  $\delta_{3 \max}^2$ . Arrow indicates shift in plotting position.

An attempt to relate  $p_i'$  fully to a practical range in  $\rho_n$ ,  $\delta_n^2$ , and  $k$  appears to be economically prohibitive because of computer cost. Rather, these results suggest that  $p_i'$  should be estimated for each specific correlation matrix and period of record.

**APPLICATION OF THE METHOD IN THE BIG LOST RIVER BASIN, IDAHO**

Frequencies of floods on the Big Lost River near the National Reactor Testing Station in southeastern Idaho are required in connection with an investigation for the Atomic Energy Commission of the potential hazard from extreme floods. The first step of the flood-frequency analysis is to correlate annual flood records in this part of Idaho with distributions that are identical save for some scaling factor.

From a report on regional flood-frequency relations (Thomas and others, 1963), 48 candidate records were selected for analysis. The coefficient of variation, square of the coefficient of skew, kurtosis, and the standard error of these moments (see Kendall and Stuart, 1963, p. 228-233) were computed for each candidate record. The records of annual floods were assumed to have distributions identical with the selected base record (Big Lost River below Mackay Reservoir) if they satisfied the following criteria:

1. The confidence limits of each moment for the record, defined by its standard error, were within the limits of the same moment for the base record.
2. The median test (Ostle, 1963, p. 473) indicated the acceptability of the hypothesis of identical distribution with the base record at the 5-percent level of significance.

These two criteria were met by 15 records (see table 1).

The next step in the analysis is to reduce the selected records to identical distributions. This is accomplished by normalizing them to the computed mean annual flood.

TABLE 1.—Name of station and period of record for floods in central Idaho having identical distributions

Name	Period of record
Beaver Creek at Spencer.....	1950-63
Lime Creek near Bennett.....	1946-56
Challis Creek near Challis.....	1944-63
Combined discharge, Big Wood River and Slough at Hailey.....	1936-57
Big Lost River below Mackay Reservoir near Mackay.....	1919-67
Surface inflow to Mackay Reservoir near Mackay.....	1919-67
Big Lost River at Howell Ranch near Chilly.....	1920-67
Salmon River at Salmon.....	1920-67
Valley Creek at Stanley.....	1921-67
Salmon River below Yankee Fork near Clayton.....	1922-67
Salmon River near Challis.....	1929-67
Big Lost River at Wild Horse near Chilly.....	1944-67
South Fork Boise River at Anderson Ranch Dam.....	1946-67
Big Wood River near Ketchum.....	1948-67
Birch Creek near Reno.....	1950-63

At this point in the analysis, the objective is to select those of the 15 concurrent records whose combination maximizes  $nk$  to realize the dual advantage of obtaining large recurrence intervals and relating outstanding events in the region to these intervals. The records for which the station years are maximized are listed in table 2.

The correlation matrix  $R$  for the six station records

$$\begin{matrix}
 & \begin{matrix} 1 & 2 & 3 & 4 & 5 & 6 \end{matrix} \\
 \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{matrix} & \begin{bmatrix}
 1 & & & & & \\
 0.379 & 1 & & & & \\
 0.418 & .941 & 1 & & & \\
 0.239 & .711 & .677 & 1 & & \\
 0.256 & .728 & .711 & .971 & 1 & \\
 0.415 & .802 & .823 & .914 & .934 & 1
 \end{bmatrix}
 \end{matrix}$$

and a  $k$  value of 46 were inputs to the computer program for which the maxima  $\{p_i'; i=1, \dots, 6\}$  were computed.

The ordered reduced variate maxima,  $p_i'$ , and the associated flood magnitude in the Big Lost River below Mackay Reservoir are listed in table 3. The flood magnitude is equal to the product of the ordered reduced variate maximum and the mean annual flood at this station. The recurrence interval,  $T$ , is  $1/p_i'$ .

The rare-flood reduced variates are plotted against their recurrence intervals in figure 8. The annual floods, in reduced variate form, are also plotted in this figure to define the relation between flood magnitude and flood frequency for the Big Lost River below Mackay Reservoir.

**DISCUSSION**

The procedure for determining the probability of occurrence of rare flood events may be extended to other kinds of dependent records, only a few alterations being required in the computer program. If the records follow a first-order Markov process as well as being cross correlated, then the simulation model is

$$X_{j+1} = AX_j + B\epsilon_{j+1},$$

where  $X_j = n \times 1$  matrix of the  $j$ th events ( $j=1, 2, \dots, k-1$ ) in  $n$  records,

$$\begin{aligned}
 A &= R_1 R_0^{-1}, \\
 BB^T &= R_0 - R_1 R_0^{-1} R_1^T, \\
 B &= E\lambda, \text{ where } E \text{ is the eigenvector matrix} \\
 &\text{and } \lambda \text{ is the diagonal matrix whose} \\
 &\text{elements are the square roots of eigen-} \\
 &\text{values,} \\
 R_0 &= \text{lag-zero correlation matrix,} \\
 R_1 &= \text{lag-one correlation matrix, and} \\
 \epsilon_j &= n \times 1 \text{ matrix of random normal numbers} \\
 &\text{with zero mean and unit variance.}
 \end{aligned}$$

(See Matalas, 1967, for more detail.)

TABLE 2.—List of annual flood records for analysis of flood-event frequency in the Big Lost River basin, Idaho, for the concurrent period 1922-67

Station	Name	Coefficient of variation		Square of skew coefficient		Kurtosis		Maximum flood	
		Value	Standard error	Value	Standard error	Value	Standard error	Cubic feet per second	Reduced variate
1	Big Lost River below Mackey Reservoir near Mackay	0.417	0.033	0.397	0.289	2.57	0.49	2,640	1.40
2	Big Lost River at Howell Ranch near Chilly	.394	.035	.170	.165	2.49	.43	4,420	2.13
3	Surface inflow to Mackay Reservoir near Mackay	.491	.052	.0493	.089	2.36	2.33	3,240	2.23
4	Valley Creek at Stanley	.474	.048	1.16	.71	3.46	1.05	2,000	2.07
5	Salmon River below Yankee Fork near Clayton	.354	.036	.217	.317	2.80	.63	10,300	1.99
6	Salmon River at Salmon	.380	.035	.153	.144	2.64	.34	16,500	1.90

TABLE 3.—Plotting positions for the relation between the rare-flood magnitude and rare-flood frequency at Big Lost River below Mackay Reservoir

Order	Exceedance probability	Recurrence interval (years)	Reduced variate maximum	Flood magnitude (cfs)
1	0.00662	151.1	2.23	4,190
2	.0115	86.9	2.13	4,000
3	.0159	62.7	2.07	3,890
4	.0218	45.8	1.99	3,750
5	.0298	33.5	1.90	3,580
6	.0432	23.2	1.40	2,640

Through special investigations for historical floods, the period of record for the maximum event may be extended from  $k$  to  $k+l$  years, where  $l$  is the period of extension. The estimation of  $p_i'$  for the  $(k+l)$ -event record may be undertaken provided the correlation matrix defined by  $k$ -event records is considered applicable to the longer period and the  $n$  maxima for the  $(k+l)$ -event records are defined.

A somewhat analogous condition to the problem of analyzing historical floods occurs when some records

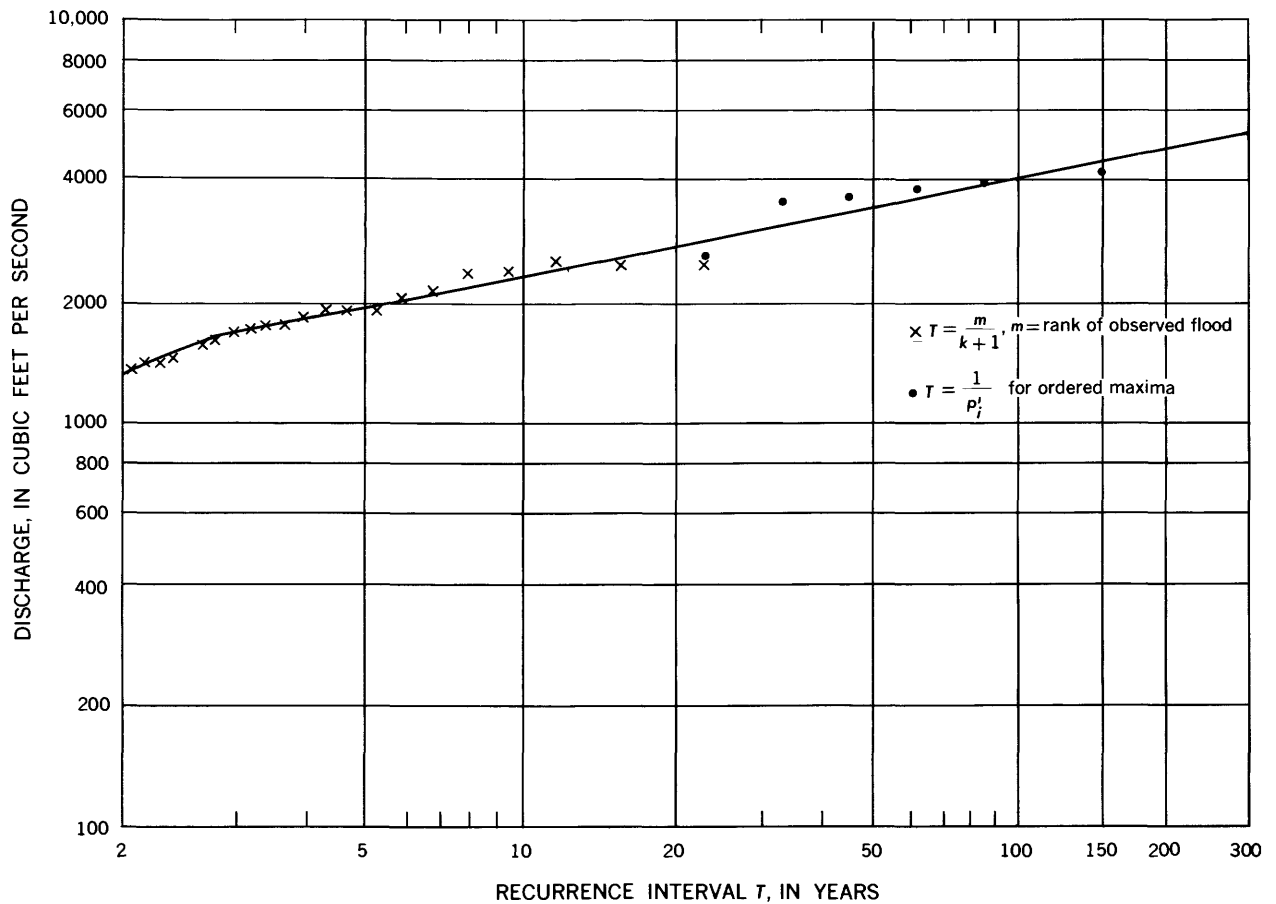


FIGURE 8.—Flood-frequency relation for Big Lost River below Mackay Reservoir based on the probability of occurrence of rare events;  $n=6$ ,  $k=46$ , and  $\bar{p}_6=0.661$ .

covering a period  $k$  have missing events. Assuming known record maxima, some uncertainty about the correlation matrix is introduced for records of this kind. If the matrix is positive semidefinite, a test of the sensitivity of  $p_i'$  to changes in individual correlation coefficients would seem appropriate. The variation in the lengths of record may be so large that the correlation matrix is inconsistent. Some suggestions for developing consistent matrices were given by Fiering (1968).

The objective in the analytical procedure for developing a regional flood-frequency relation is to maximize the number of station years (increasing the sample size) which tends to minimize  $p_i'$ . Because  $p_i'$  varies almost directly with  $1/nk$ , the objective can be attained by seeking a large number of short records or a few long records. The objection to the use of short records is that the sample may be nonrepresentative in comparison with norms established by much longer sequences of events. Unless statistical comparisons between long and short records indicate non-homogeneity, no reason exists for rejecting short records. Sampling errors will be larger for short records, but this disadvantage may be outweighed by the substantially larger values of  $nk$  that will be provided by the shorter than the longer records.

### CONCLUSIONS

The technique of regional flood-frequency analysis that has been introduced takes advantage of the random variation of flood intensities in both time and space. Previous methods of analysis are virtually limited to the consideration of variation in time. A result of applying this new technique is to increase the recurrence interval for the maximum flood event; because of correlation between records, it is unlikely that this recurrence interval would approach the number of station-years of record available.

Using Monte Carlo simulation techniques, the exceedance probability for rare flood events may be estimated by the simultaneous consideration of concurrent records of annual extremes which are dependent and identically distributed. The method for obtaining rare-event probabilities is to generate sets of normal multivariates equal in number and duration to the observed sets of records through the simulation model

$$X = B\epsilon$$

to order the simulated record maxima and to compute the normal exceedance probabilities of the ordered maxima. This scheme of generating synthetic variates preserves the observed dependence in records.

In applying the procedure of estimating exceedance probabilities to regional flood-frequency problems it is assumed that a means of defining the region of identical distributions is at hand, that the estimate of exceedance probability is independent of the underlying statistical distribution of observed floods, and that the sample correlation matrix is a reasonable estimate of the true matrix. Indications are that the probability estimates are independent of the distribution of observed floods.

If the correlation coefficients for the records are all equal, the rare-event probabilities vary in a predictable manner with changes in numbers of records, period of record, and magnitude of the coefficient. If the correlation matrix is a mixture of coefficients, variations in the probabilities are a function not only of an equivalent mean coefficient, the number of records, and the period of record, but also of the deviation of the coefficients about the mean coefficient. Even though the probabilities associated with a general correlation matrix also appear to vary predictably with these variables, any general solution would not seem economically justified because of excessive computer costs (about  $2nk$  seconds per run). It would seem to be preferable in the interest of economy to estimate the rare-event exceedance probabilities for each set of data under analysis.

A set of criteria for selecting records having identical distributions has been identified in this report. Other sets of criteria for making this selection are possible, and the most definitive must be sought. The simulation program is equally applicable to many kinds of rare hydrologic and hydrometeorologic events such as precipitation, floods, and temperatures.

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COMPUTER PROGRAM

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FLOOD-FREQUENCY RELATION BASED ON REGIONAL RECORD MAXIMA

F13

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C      SUBROUTINES AND FUNCTION SUBPROGRAMS
C      MTRXIN
C      MTRXOT
C      EGN
C      GMPRD
C      MAX
C      MIN
C      LISD
C
C      METHOD
C      REFER TO P.H.CARRIGAN, JR., 1971, A FLOOD FREQUENCY RELATION
C      BASED ON REGIONAL RECORD MAXIMA, U.S.GEOL.SURVEY PROF.
C      PAPER 434-F
C
C      REMARKS
C      EXECUTION TIME ABOUT 2NK SECONDS PER DATA SET. FORT PLUS
C      LKED TIME LESS THAN 15 SECONDS
C      MAXIMUM LINES OF OUTPUT = 800 + 205*(NO. OF DATA SETS)
C
C      DIMENSION X(20,100),Y(20),P(20),JUNK(20) #1.0000
C      DOUBLE PRECISION D1,D2,D3,D4,D5,D6,U,RO(210),R(400),B(400),B1(400) #1.0010
C      DOUBLE PRECISION X1(20),E(20),RBC(210) #1.0020
C      DATA D1,D2,D3/0.049867347,0.0211410061,0.0032776263/ #1.0030
C      DATA D4,D5,D6/0.0000380036,0.0000488906,0.0000053830/ #1.0040
101  FORMAT(I4) #1.0050
102  FORMAT(2I4) #1.0060
201  FORMAT('1',12X,'EMPIRICAL EXCEEDANCE PROBABILITIES'/13X,'FOR ORDER#1.0070
      1ED SET OF ',I2,'-RECORD MAXIMA'/18X,'WITH',I4,' EVENTS PER RECORD'#1.0080
      2////) #1.0090
202  FORMAT(' ',20X,'CORRELATION MATRIX'//) #1.0100
203  FORMAT('0/' ' ',20X,'SQRT EIGENVALUE MATRIX'//) #1.0110
204  FORMAT('0/' ' ',20X,'EIGENVECTOR MATRIX'//) #1.0120
205  FORMAT('0',10X, 'THE LFAST NUMBER OF HIGH ORDER SIGNIFICANT DIGIT#1.0130
      1S (NSD) IN A CHCK OF CHARACTERISTIC EQUATION IS',I3,'.'//) #1.0140
206  FORMAT('0',9X,'ORDER',8X,8(I2,10X)) #1.0150
207  FORMAT(4X,'PROBABILITY',5X,8(F9.6,3X)) #1.0160
208  FORMAT('0',10X,'INCORRECT NUMBER OF DATA CARDS FOR CORRELATION MAT#1.0170
      1RIX'//) #1.0180
209  FORMAT('0',10X,'MATRIX OF CORRELATION COEFFICIENTS DOES NOT TEST T#1.0190
      10 BE POSITIVE SEMIDEFINITE. (DETERMINANT LESS THAN ZERO.)'/ #1.0200
      2' ',10X,'THIS MAY BE A RESULT OF INACCURACIES IN THE COMPUTER ALGO#1.0210
      3RITHMS.'/' ' ',10X,'***THIS PROGRAM WILL NOT OPERATE ON THIS DATA SE#1.0220
      4T BEYOND STATEMENT 70***'). #1.0230
C
C      RFAD(5,101)IX #1.0240
1  READ(5,102,END=500)N,K #1.0250
C
C      READ IN RO
C
C      CALL MTRXIN(RO,N,IER) #1.0260
C      WRITE(6,201)N,K #1.0270
C
C      CHECK FOR ERROR IN INPUT
C
C      IF(IER)60,61,60 #1.0280
60  WRITE(6,208) #1.0290
      GO TO 1 #1.0300
61  WRITE(6,202) #1.0310
C
C      PRINT CORRELATION MATRIX
C
C      CALL MTRXOT(RO,N,1) #1.0320
C
C      COMPUTE EIGENVECTORS AND EIGENVALUES OF RO
C
C      NSD=0 #1.0330
C      CALL EGN(RO,R,RBC,N,NSD) #1.0340
C
C      CHECK FOR POSITIVE SEMIDEFINITE CORRELATION MATRIX
C

```

```

DO 70 I=1,N                                     #1.0350
IJ=(I*I+I)/2                                    #1.0360
IF(RO(IJ).GE.0.0) GO TO 70                       #1.0370
WRITE(6,209)                                     #1.0380
GO TO 71                                          #1.0390
70 CONTINUE                                       #1.0400
C
C   COMPUTE SQUAREROOTS OF EIGENVALUFS AND SET OFF-DIAGONAL ELEMENTS
C   OF RO TO ZERO. PUT ELEMENTS IN GENERAL STORAGE MODE.
C
DO 2 I=1,N                                       #1.0410
DO 2 J=1,N                                       #1.0420
IF(I-J)5,6,6                                     #1.0430
5 IJ=I+(J*J-J)/2                                #1.0440
GO TO 3                                          #1.0450
6 IJ=J+(I*I-I)/2                                #1.0460
IF(I.NE.J) GO TO 3                              #1.0470
RO(IJ)=DSQRT(RO(IJ))                            #1.0480
GO TO 4                                          #1.0490
3 RO(IJ)=0.0                                     #1.0500
4 K1=N*(J-1)+I                                  #1.0510
2 B1(K1)=RO(IJ)                                 #1.0520
C
C   PRINT SQRT OF EIGENVALUFS AND EIGENVECTOR MATRICES
C
WRITE(6,203)                                     #1.0530
CALL MTRXOT(B1,N,0)                             #1.0540
WRITE(6,204)                                     #1.0550
CALL MTRXOT(R,N,0)                              #1.0560
C
C   PRINT SIGNIFICANT DIGITS MESSAGE
C
IF(NSD.EQ.0) GO TO 73                           #1.0570
72 WRITE(6,205)NSD                              #1.0580
C
C   COMPUTE PRINCIPAL COMPONENT MATRIX B
C
73 CALL GMPRD(R,B1,B,N,N,N)                     #1.0590
C
C   INITIALIZE P(J) TO ZERO
C
DO 7 J=1,N                                       #1.0600
7 P(J)=0.0                                       #1.0610
C
C   IN THE 1000 ITERATION LOOP ENDING IN STATEMENT 10, AN N BY K
C   RECORD OF X(J,KK) IS SIMULATED (USING MATRIX B AND A NORMAL RANDOM
C   NUMBER E), N COLUMN MAXIMA Y(J) ARE SELECTED FROM X(J,KK),
C   Y(J) ARE PLACED IN DESCENDING ORDER (J=1,2,...,N), NORMAL
C   EXCEEDANCE PROBABILITY C FOR EACH Y(J) IS FOUND, AND THE 1000
C   ITEM SUM OF C FOR EACH J, P(J), IS COMPUTED.
C
DO 10 M=1,1000                                  #1.0620
C
C   GENERATE X(J,KK) ARRAY
C
DO 11 KK=1,K                                    #1.0630
DO 15 I=1,N                                    #1.0640
C
C   SUBROUTINE GAUSS IS IMBEDDED IN MAIN PROGRAM BY FOLLOWING
C   10 STATEMENTS
C
AA=0.0                                           #1.0650
DO 12 II=1,12                                   #1.0660
IY=IX*65539                                       #1.0670
IF(IY)13,14,14                                   #1.0680
13 IY=IY+1+2147483647                             #1.0690
14 YFL=IY                                         #1.0700
YFL=YFL*0.4656613E-9                             #1.0710
IX=IY                                             #1.0720
12 AA=AA+YFL                                       #1.0730
15 E(I)=AA-6.0                                     #1.0740
C

```

```

C      FIND K VECTOR PRODUCTS OF X1 - X(N,K)
C
      CALL GMPRD(B,E,X1,N,N,1)                #1.0750
      DO 11 I=1,N                             #1.0760
      J=I                                     #1.0770
11     X(J,KK)=X1(I)                         #1.0780
C
C      FIND COLUMN MAXIMA
C
      DO 20 J=1,N                             #1.0790
      D=-1000.0                               #1.0800
      DO 21 KK=1,K                           #1.0810
      CM=AMAX1(X(J,KK),D)                   #1.0820
21     D=CM                                  #1.0830
20     Y(J)=D                               #1.0840
C
C      ORDER COLUMN MAXIMA
C
      NN=N-1                                  #1.0850
30     DO 31 J=1,NN                         #1.0860
      IF((Y(J)-Y(J+1)).GE.0.0) GO TO 31      #1.0870
      A=Y(J+1)                               #1.0880
      Y(J+1)=Y(J)                           #1.0890
      Y(J)=A                                 #1.0900
31     CONTINUE                             #1.0910
      NN=NN-1                                #1.0920
      IF(NN.GT.0) GO TO 30                  #1.0930
C
C      COMPUTE NORMAL EXCEEDANCE PROBABILITY C OF EACH COLUMN MAXIMUM
C      Y(J). (REF. - EQN. 26.2.19 NBS HANDBK (1965))
C
      DO 10 J=1,N                             #1.0940
      IF(Y(J).GE.0.0) GO TO 40              #1.0950
      U=-Y(J)                                #1.0960
      GO TO 41                               #1.0970
40     U=Y(J)                                #1.0980
41     IF(U.GT.5.0) GO TO 42                #1.0990
      IF(U.LT.0.0001) GO TO 43              #1.1000
      G=1.0+D1*U+D2*U*U+D3*U*U*U+D4*U**4+D5*U**5+D6*U**6 #1.1010
      C=0.5*G**(-16)                       #1.1020
      GO TO 44                               #1.1030
42     C=0.0                                #1.1040
      GO TO 44                               #1.1050
43     C=0.5                                #1.1060
44     IF(Y(J).GE.0.0) GO TO 10            #1.1070
      C=C-1.0                               #1.1080
C
C      SUMMATION OF C = P(J)
C
10     P(J)=P(J)+C                          #1.1090
C
C      COMPUTE MEAN EXCEEDANCE PROBABILITY P(J) FOR JTH-ORDER MAXIMUM
C
      DO 45 J=1,N                             #1.1100
45     P(J)=P(J)/1000.0                   #1.1110
C
C      PRINT P(J)
C
      JS=1                                    #1.1120
      IF(N.LE.8) GO TO 50                  #1.1130
      JE=8                                  #1.1140
      GO TO 51                             #1.1150
50     JE=N                                #1.1160
51     DO 52 J=JS,JE                       #1.1170
52     JUNK(J)=J                           #1.1180
      WRITE(6,206)(JUNK(J),J=JS,JE)        #1.1190
      WRITE(6,207)(P(J),J=JS,JE)          #1.1200
      JS=JS+8                              #1.1210
      IF(JS.GT.N) GO TO 53                 #1.1220
      JE=JE+8                              #1.1230

```

## STATISTICAL STUDIES IN HYDROLOGY

```

IF(LE.LT.N) GO TO 51 #1.1240
JE=N #1.1250
GO TO 51 #1.1260
53 CONTINUE #1.1270
71 GO TO 1 #1.1280
500 STOP #1.1290
END #1.1300
SUBROUTINE MTRXIN(A,NORD,IER) #1.1310
C
C PURPOSE
C READS DATA ELEMENTS OF SYMMETRIC MATRIX FROM CARDS
C
C DESCRIPTION OF VARIABLES
C A - INPUT MATRIX
C NORD - ORDER OF INPUT MATRIX. MUST BE 20 OR LESS.
C IER - ERROR IN DATA CARDS CODE
C IER = 0 NO ERROR
C = 1 INCORRECT NUMBER OF CARDS
C
C LISTING OF DATA CARDS
C FORMAT IS 7F10.0. COLUMNS 71-80 FOR IDENTIFICATION. DATA
C ELEMENTS PUNCHED BY ROW STARTING WITH DIAGONAL ELEMENT AND
C PROCEEDING TO THE RIGHT IN THAT ROW, WITH LISTING CONTINUED
C TO NEW CARDS IF NECESSARY. LAST CARD FOR LAST ROW IN MATRIX
C FOLLOWED BY A CARD WITH 9 PUNCHED IN COLUMN 1.
C
C REFERENCE
C ADAPTATION OF SUBROUTINE MATIN. REFER TO INTERNATIONAL
C BUSINESS MACHINES, 1968, SYSTEM/360 SCIENTIFIC SUBROUTINE
C PACKAGE (360-CM-03X) VERSION III, PROGRAMMER'S MANUAL.
C WHITE PLAINS, NEW YORK, INTERNATIONAL BUSINESS MACHINES,
" P.453
C
DOUBLE PRECISION A(1),CARD(8) #1.1320
1 FORMAT(7F10.0) #1.1330
IER=0 #1.1340
ICOLT=NORD #1.1350
IROCR=1 #1.1360
C
C COMPUTE NUMBFR OF CARDS IN ROW
C
2 IRCDS=(ICOLT-1)/7+1 #1.1370
C
C SET UP LOOP FOR NUMBER OF CARDS IN ROW
C
DO 3 K=1,IRCDS #1.1380
READ(5,1)(CARD(I),I=1,7) #1.1390
L=0 #1.1400
C
C COMPUTE COLUMN NUMBER FOR FIRST FIELD IN CURRENT CARD
C
JS=(K-1)*7+NORD-ICOLT+1 #1.1410
JE=JS+6 #1.1420
C
C SET UP LOOP FOR DATA ELEMENTS WITHIN CARD
C
DO 4 J=JS,JE #1.1430
IF(J-NORD)5,5,3 #1.1440
5 IF(IROCR-J)6,7,7 #1.1450
6 IJ=IROCR+(J*J-J)/2 #1.1460
GO TO 8 #1.1470
7 IJ=J+(IROCR*IROCR-IROCR)/2 #1.1480
8 L=L+1 #1.1490
4 A(IJ)=CARD(L) #1.1500
3 CONTINUE #1.1510
IROCR=IROCR+1 #1.1520
IF(NORD-IROCR)9,10,10 #1.1530
10 ICOLT=ICOLT-1 #1.1540
GO TO 2 #1.1550

```

9	READ(5,1)CARD(1)	#1.1560
	IF(CARD(1)-9.D9)11,12,11	#1.1570
11	IER=1	#1.1580
12	CONTINUE	#1.1590
	RETURN	#1.1600
	END	#1.1610
	SUBROUTINE MTRXOT(A,NORD,IS)	#1.1620
C	PURPOSE	
C	PRINTOUT LISTING OF ANY SIZED MATRIX	
C		
C	DESCRIPTION OF VARIABLES	
C	A - OUTPUT MATRIX	
C	NORD - ORDER OF MATRIX	
C	IS - STORAGE MODE OF A	
C	IS = 0 GENERAL	
C	IS = 1 SYMMETRIC	
C	IS = 2 DIAGONAL	
C		
C	SUBROUTINES REQUIRED	
C	LOC	
C		
C	REFERENCE	
C	ADAPTATION OF SUBROUTINE MXOUT. REFER TO INTERNATIONAL	
C	BUSINESS MACHINES, 1968, SYSTEM/360 SCIENTIFIC SUBROUTINE	
C	PACKAGE (360-CM-03X) VERSION III, PROGRAMMER'S MANUAL.	
C	WHITE PLAINS, NEW YORK, INTERNATIONAL BUSINESS MACHINES,	
C	P.454	
	DOUBLE PRECISION A(1),B(8)	#1.1630
201	FORMAT(20X,7(3X,I3,10X))	#1.1640
202	FORMAT('0',11X,I3,7(D16.6))	#1.1650
	J=1	#1.1660
1	LS=1	#1.1670
2	JNT=J+6	#1.1680
	IF(JNT-NORD)4,4,3	#1.1690
3	JNT=NORD	#1.1700
C		
C	NUMBER COLUMNS	
C		
4	WRITE(6,201)(JCUR,JCUR=J,JNT)	#1.1710
	LE=LS+57	#1.1720
	DO 5 L=LS,LE	#1.1730
C		
C	PRINT LINE	
C		
	DO 6 K=1,7	#1.1740
	KK=K	#1.1750
	JT=J+K-1	#1.1760
	CALL LOC(L,JT,IJ,NORD,NORD,IS)	#1.1770
	B(K)=0.0	#1.1780
	IF(IJ)8,8,7	#1.1790
7	B(K)=A(IJ)	#1.1800
C		
C	IF LAST COLUMN, WRITE ROW'S NUMBER AND ELEMENTS	
C		
8	IF(JT-NORD)6,9,9	#1.1810
6	CONTINUE	#1.1820
9	WRITE(6,202)L,(B(JW),JW=1,KK)	#1.1830
C		
C	IF END OF ROWS, CHECK FOR MORE COLUMNS	
C		
	IF(NORD-L)10,10,5	#1.1840
5	CONTINUE	#1.1850
C		
C	CHECK FOR MORE OUTPUT	
C		
	LS=LS+58	#1.1860
	GO TO 2	#1.1870
		#1.1880



```

C      COMPUTE INITIAL AND FINAL NORMS, ANORM AND ANORMX
C
5  ANORM=0.0D0                                #1.0260
   DO 7 I=1,N                                  #1.0270
   DO 7 J=I,N                                  #1.0280
   IF (I-J) 6,7,6                             #1.0290
6  IA=I+(J*J-J)/2                             #1.0300
   ANORM=ANORM+A(IA)*A(IA)                   #1.0310
7  CONTINUE                                    #1.0320
   IF (ANORM) 33,33,8                         #1.0330
8  ANORM=DSQRT(2*ANORM)                       #1.0340
   ANRMX=ANORM*RANGE/DFLOAT(N)              #1.0350
C
C      INITIALIZE INDICATORS AND COMPUTE THRESHOLD, THR
C
   IND=0                                       #1.0360
   THR=ANORM                                  #1.0370
9  THR=THR/DFLOAT(N)                          #1.0380
10 L=1                                        #1.0390
11 M=L+1                                       #1.0400
C
C      COMPUTE SIN AND COS
C
12 MQ=(M*M-M)/2                               #1.0410
   LQ=(L*L-L)/2                               #1.0420
   LM=L+MQ                                    #1.0430
   IF (DABS(A(LM))-THR) 26,13,13             #1.0440
13 IND=1                                       #1.0450
   LL=L+LQ                                    #1.0460
   MM=M+MQ                                    #1.0470
   Z=A(LL)                                    #1.0480
   Q=A(MM)                                    #1.0490
   X=(Z-Q)/2                                  #1.0500
   Y=-A(LM)/DSQRT(A(LM)*A(LM)+X*X)          #1.0510
   IF (X) 14,15,15                           #1.0520
14 Y=-Y                                       #1.0530
15 SINX=Y/DSQRT(2*(1+(DSQRT(1-Y*Y))))        #1.0540
   SINX2=SINX*SINX                          #1.0550
   COSX2=1-SINX2                             #1.0560
   COSX=DSQRT(COSX2)                        #1.0570
   SINCS=SINX*COSX                          #1.0580
C
C      ROTATE L AND M COLUMNS
C
   ILQ=N*(L-1)                               #1.0590
   IMQ=N*(M-1)                               #1.0600
   DO 25 I=1,N                                #1.0610
   IQ=(I*I-I)/2                              #1.0620
   IF (I-L) 16,23,16                         #1.0630
16 IF (I-M) 17,23,18                         #1.0640
17 IM=I+MQ                                    #1.0650
   GO TO 19                                   #1.0660
18 IM=M+IQ                                    #1.0670
19 IF (I-L) 20,21,21                         #1.0680
20 IL=I+LQ                                    #1.0690
   GO TO 22                                   #1.0700
21 IL=L+IQ                                    #1.0710
22 X=A(IL)*COSX                              #1.0720
   W=A(IM)*SINX                              #1.0730
   Y=A(IL)*SINX                              #1.0740
   W1=-A(IM)*COSX                           #1.0750
   A(IM)=Y-W1                                #1.0760
   A(IL)=X-W                                  #1.0770
23 IF (MV-1) 24,25,24                       #1.0780
24 ILR=ILQ+I                                 #1.0790
   IMR=IMQ+I                                 #1.0800
   Z=R(ILR)*COSX                             #1.0810
   Q=R(IMR)*SINX                             #1.0820
   X=Z-Q                                      #1.0830
   Z=R(ILR)*SINX                             #1.0840

```



```

Q=-R(IMR)*COSX                                     #1.0850
R(IMR)=Z-Q                                          #1.0860
R(ILR)=X                                            #1.0870
25 CONTINUE                                         #1.0880
X=2*A(LM)*SINCS                                    #1.0890
Z=A(LL)*COSX2                                       #1.0900
Q=-A(MM)*SINX2                                      #1.0910
Y=Z-Q                                               #1.0920
Z=A(LL)*SINX2                                       #1.0930
Q=-A(MM)*COSX2                                      #1.0940
W=Z-Q                                               #1.0950
W1=-W                                               #1.0960
A(LM)=O                                              #1.0970
A(LL)=Y-X                                           #1.0980
A(MM)=W+X                                           #1.0990

C
C   TESTS FOR COMPLETION
C
C   TEST FOR M = LAST COLUMN
C
26 IF (M-N) 27,28,27                                #1.1000
27 M=M+1                                             #1.1010
   GO TO 12                                          #1.1020

C
C   TEST FOR L = SECOND FROM LAST COLUMN
C
28 IF (L-(N-1)) 29,30,29                            #1.1030
29 L=L+1                                             #1.1040
   GO TO 11                                          #1.1050
30 IF (IND-1) 32,31,32                              #1.1060
31 IND=0                                             #1.1070
   GO TO 10                                          #1.1080

C
C   COMPARE THRESHOLD WITH FINAL NORM
C
32 IF (THR-ANRMX) 33,33,9                            #1.1090

C
C   SORT EIGENVALUES AND EIGENVECTORS
C
33 IQ=-N                                             #1.1100
   DO 37 I=1,N                                       #1.1110
     IQ=IQ+N                                         #1.1120
     LL=I+(I*I-I)/2                                  #1.1130
     JQ=N*(I-2)                                       #1.1140
     DO 37 J=I,N                                       #1.1150
       JQ=JQ+N                                       #1.1160
       MM=J+(J*J-J)/2                                 #1.1170
       IF (A(LL)-A(MM)) 34,37,37                    #1.1180
34 X=A(LL)                                           #1.1190
   A(LL)=A(MM)                                       #1.1200
   A(MM)=X                                           #1.1210
   IF (MV-1) 35,37,35                                #1.1220
35 DO 36 K=1,N                                       #1.1230
   ILR=IQ+K                                          #1.1240
   IMR=JQ+K                                          #1.1250
   X=R(ILR)                                          #1.1260
   R(ILR)=R(IMR)                                     #1.1270
36 R(IMR)=X                                          #1.1280
37 CONTINUE                                         #1.1290

C
C   DETERMINE LEAST NUMBER OF SIGNIFICANT DIGITS
C
   IF (MV-1) 38,48,38                                #1.1300
38 CONTINUE                                         #1.1310

C
C   COMPUTE MATRIX ELEMENTS OF PRODUCT OF ORIGINAL MATRIX AND CHARAC-
C   TERISTIC VECTORS, X, AND OF EIGENVALUES AND CHARACTERISTIC
C   VECTORS, Z.
C

```

FLOOD-FREQUENCY RELATION BASED ON REGIONAL RECORD MAXIMA

F21

```

DO 45 I2=1,N
Y=A(I2*(I2+1)/2)
I1=N*(I2-1)
DO 43 L=1,N
X=0
DO 42 M=1,N
IF (M-L) 39,40,40
39 I=L*(L-1)/2+M
GO TO 41
40 I=M*(M-1)/2+L
41 K=I1+M
42 X=X+B(I)*R(K)
Z=Y*R(I1+L)
#1.1320
#1.1330
#1.1340
#1.1350
#1.1360
#1.1370
#1.1380
#1.1390
#1.1400
#1.1410
#1.1420
#1.1430
#1.1440

C
C DETERMINE LEAST NUMBER OF SIGNIFICANT DIGITS IN X OR Y FOR EACH
C ELEMENT OF MATRIX
C
NSD=LISD(X,Z,15)
IF(NSD)43,49,43
49 I1L=I1+L
WRITE(6,201)I1L,X,Z
#1.1450
#1.1460
#1.1470
#1.1480
43 MSD=MIN(NSD,MSD)
MSDR=MIN(MSD,MSDR)
K1=I2*(I2+1)/2+1
IF (I2-N) 44,46,46
#1.1490
#1.1500
#1.1510
#1.1520
44 A(K1)=MSD
#1.1530
45 MSD=15
#1.1540
46 DO 47 I=1,N
K=I*(I+1)/2
L=K+1
#1.1550
#1.1560
#1.1570
47 B(K)=A(L)
#1.1580
B(NX2)=MSD
#1.1590
NSD=MSDR
#1.1600
48 CONTINUE
#1.1610
RETURN
#1.1620
END
#1.1630
FUNCTION LISD(X,Y,N)
#1.1640
PURPOSE
TO FIND SMALLER NUMBER OF SIGNIFICANT DIGITS IN TWO NUMBERS
X AND Y, IF NUMBER IS LESS THAN N
REMARKS
PROGRAMMED BY MARSHALL HELLMAN, COMPUTER CENTER DIVISION,
U.S. GEOLOGICAL SURVEY
DOUBLE PRECISION X,Y,EPS,T
LISD=0
IF(DABS(Y).GE.DABS(X)) GO TO 1
T=X
IF(T.EQ.0.0) GO TO 2
GO TO 3
1 T=Y
IF(T.EQ.0.0) GO TO 2
3 EPS=DLNG10(DABS(T))
IX=EPS
IF(EPS.GE.0.0) GO TO 4
IF(EPS.GE.IX) GO TO 4
IX=IX-1
4 IF(Y.NE.X) GO TO 5
2 LISD=N
GO TO 6
5 T=DLNG10(DABS(Y-X))
IT=T
IF(T.GE.0.0) GO TO 7
IF(T.GE.IT) GO TO 7
IT=IT-1
7 LISD=IX-IT-1
LISD=MAX(LISD,0)
6 RETURN
#1.1650
#1.1660
#1.1670
#1.1680
#1.1690
#1.1700
#1.1710
#1.1720
#1.1730
#1.1740
#1.1750
#1.1760
#1.1770
#1.1780
#1.1790
#1.1800
#1.1810
#1.1820
#1.1830
#1.1840
#1.1850
#1.1860
#1.1870
#1.1880
#1.1890
END

```

## STATISTICAL STUDIES IN HYDROLOGY

```

C      FUNCTION MAX(M,N)                                     #1.1900
C      PURPOSE
C      TO DETERMINE IF INTEGER M IS LARGER THAN SELECTED INTEGER N
C
C      REMARKS
C      PROGRAMMED BY MARSHALL HELLMAN, COMPUTER CENTER DIVISION,
C      U.S. GEOLOGICAL SURVEY
C
C      IF(M.GT.N) GO TO 1                                     #1.1910
C      MAX=N                                                 #1.1920
C      RETURN                                               #1.1930
1     MAX=M                                                 #1.1940
C      RETURN                                               #1.1950
C      END                                                 #1.1960
C      FUNCTION MIN(M,N)                                     #1.1970
C      PURPOSE
C      TO DETERMINE IF INTEGER M IS SMALLER THAN SELECTED INTEGER N
C
C      REMARKS
C      PROGRAMMED BY MARSHALL HELLMAN, COMPUTER CENTER DIVISION,
C      U.S. GEOLOGICAL SURVEY
C
C      IF(M.LE.N) GO TO 1                                     #1.1980
C      MIN=N                                                 #1.1990
C      RETURN                                               #1.2000
1     MIN=M                                                 #1.2010
C      RETURN                                               #1.2020
C      END                                                 #1.2030
C      SUBROUTINE GMPRD(A,B,R,N,M,L)                         #1.2040
C
C      PURPOSE
C      PRODUCT OF TWO GENERAL-MODE-STORAGE MATRICES
C      DESCRIPTION OF VARIABLES
C      A - FIRST INPUT MATRIX, N ROWS, M COLUMNS
C      B - SECOND INPUT MATRIX, M ROWS, L COLUMNS
C      R - OUTPUT MATRIX, N ROWS, L COLUMNS
C
C      REFERENCE
C      INTERNATIONAL BUSINESS MACHINES, 1968, SYSTEM/360 SCIENTIFIC
C      SUBROUTINE PACKAGE (360-CM-03X) VERSION III, PROGRAMMER'S
C      MANUAL. WHITE PLAINS, NEW YORK, INTERNATIONAL BUSINESS
C      MACHINES, P.99
C
C      DOUBLE PRECISION A(1),B(1),R(1)                       #1.2050
C      IR=0                                                  #1.2060
C      IK=-M                                                 #1.2070
C      DO 1 K=1,L                                           #1.2080
C      IK=IK+M                                              #1.2090
C      DO 1 J=1,N                                           #1.2100
C      IR=IR+1                                              #1.2110
C      JI=J-N                                               #1.2120
C      IB=IK                                                #1.2130
C      R(IR)=0.0                                           #1.2140
C      DO 1 I=1,M                                           #1.2150
C      JI=JI+N                                              #1.2160
C      IB=IB+1                                              #1.2170
1     R(IR)=R(IR)+A(JI)*B(IB)                               #1.2180
C      RETURN                                               #1.2190
C      END

```





