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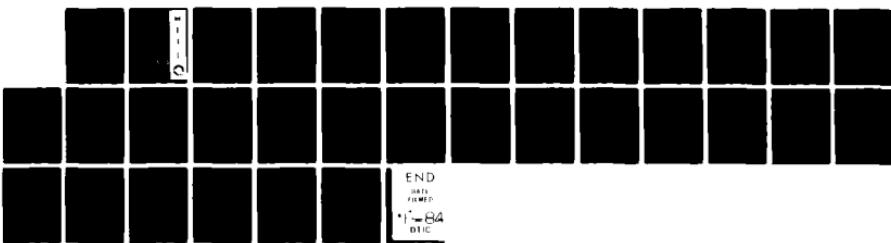
APPLICATION OF A FEATURE SELECTION TECHNIQUE TO SAMPLES
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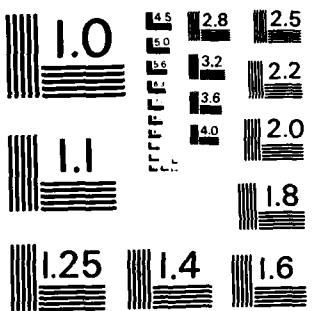
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**Application of a feature
selection technique to samples
of high resolution synthetic
aperture radar imagery**

Richard A. Hevenor

July 1983

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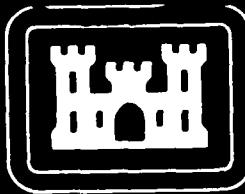
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20. ABSTRACT (Continue on reverse side if necessary and identify by block number) A feature selection technique was applied to samples of synthetic aperture radar imagery. This technique was applied to four classes of terrain features on selected samples of radar imagery. The four classes considered were forests, cities, agricultural fields, and water. A feature vector was computed from samples of each class. A linear transformation was utilized to develop a new feature vector of reduced dimensionality. This transformation chooses those features that are most effective for performing class separability.		

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PREFACE

The authority for performing the work described in this research note is contained in Project 4A161102B52C, "Research in Geodetic, Cartographic, and Geographic Sciences."

The work described in this research note represents an application of a standard feature selection technique to samples of high resolution synthetic aperture radar imagery. The task was performed under the supervision of Dr. Frederick W. Rohde, Team Leader, Center for Physical Sciences and Mr. Melvin Crowell, Jr., Director, Research Institute.

COL Edward K. Wintz, CE, was the Commander and Director and Mr. Robert P. Macchia was the Technical Director of the Engineer Topographic Laboratories during the study period.

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APPLICATION OF A FEATURE SELECTION TECHNIQUE TO SAMPLES OF HIGH RESOLUTION SYNTHETIC APERTURE RADAR IMAGERY

INTRODUCTION The purpose of this research note is to show the application of a feature selection technique to samples of synthetic aperture radar imagery and to present some preliminary results. In the past, feature selection techniques have been applied to data computed from digitized aerial photography. However, it appears that no one has as yet applied feature selection techniques to high resolution radar imagery. In order to perform classification of terrain features using radar imagery, feature selection is an important initial step. Feature selection consists of choosing those features that are most effective for showing class separability and for performing a reduction in the dimensionality of the feature vector. The following sections will present a discussion of the feature selection technique, along with its application to selected samples of radar imagery.

METHODOLOGY The application of pattern recognition techniques is accomplished usually in two steps, namely, feature selection and classifier design. The feature selection process that precedes the classification process consists of techniques applicable to one class or to multiple classes. The feature selection technique to be discussed here is applicable to multiple classes. It provides the capability of reducing the number of components of the original feature vector in such a way that the resulting components are optimized to show class separability. The feature selection technique comes from the field of discriminant analysis of statistics and is independent of the probability density functions of the feature vector data. The feature selection operation can be expressed as a linear transformation of the following form:

$$Y = AX \quad (1)$$

where X is the original feature vector with dimensionality $n \times 1$; A is the transformation matrix of dimensionality $m \times n$, where m is less than n ; and Y is the transformed feature vector with dimensionality $m \times 1$. The feature selection problem is now reduced to determining the matrix A . In order to calculate A , use is made of the within-class and between-class scatter matrices. A within-class scatter matrix shows the scatter of samples around their class expected vector and can be expressed by

$$S_w = \sum_{i=1}^N P(\omega_i) C_i \quad (2)$$

where S_w is the within-class scatter matrix, $P(\omega_i)$ is the a priori probability of the i^{th} class, C_i is the covariance matrix of the i^{th} class, and N is the total number of classes. A between-class scatter matrix can be defined in many ways; however, the following definition was the one utilized here:

$$S_b = C_1 + C_2 + (M_1 - M_2)(M_1 - M_2)^T \quad (3)$$

where S_b is the between-class scatter matrix, C_1 is the covariance matrix for class 1, C_2 is the covariance matrix for class 2, M_1 is the mean vector for class 1, M_2 is the mean vector for class 2, and T means transpose. This definition of the between-class scatter matrix is valid only for the case where N (the number of classes) is equal to two. In order to have criteria for class separability, a number must be derived from the scatter matrices. This number should increase when the distances between points belonging to different classes are increasing or when the distances between points belonging to the same class are decreasing. One criterion is the use of J_1 , which can be defined as follows:

$$J_1 = \text{trace}(S_{2m}^{-1} S_{1m}) \quad (4)$$

where

$$S_{2m} = AS_w A^T \text{ and } S_{1m} = AS_b A^T$$

The feature selection problem now requires that we select the particular transformation matrix A , which maximizes the value of J_1 . Fukunaga¹ shows that A is made up of the normalized eigenvectors of the matrix $S_w^{-1} S_b$.

$$A^T = [\phi_1 \ \phi_2 \ \dots \ \phi_m] \quad (5)$$

where ϕ_1 is the eigenvector associated with the largest eigenvalue, ϕ_2 is the eigenvector associated with the second largest eigenvalue, etc. Once the matrix A is computed from (5), the new feature vector Y can be computed for each point in each class.

INVESTIGATION

The feature selection technique was applied to samples of high resolution synthetic aperture radar imagery taken over the Huntsville, Alabama, area with the APD-10 radar system. Sections of the radar imagery were digitized and stored on a digital disk unit. A Lexidata system 3400 display processor was used to display the images on a cathode ray tube and to take 100 samples for each of four terrain classes from the imagery. Each sample consisted of a 32 by 32 pixel element window located within a section of one particular terrain class. The four classes considered were cities (combination of commercial and residential structures, DLMS category #504 FIC 301 and #505 FIC 401), fields (agriculture used primarily for crop and pasture land, DLMS category #510 FIC 950), water (rivers with smooth fresh water, DLMS category #510 FIC 940) and fresh water subject to ice (lakes and reservoirs, DLMS category #510 FIC 943), and forests (mixed trees, deciduous and evergreens, DLMS category #510 FIC 954). A feature vector consisting of 13 components was computed for each sample. These 13 components were made up of the first- and second-order gray level histogram statistics computed from each sample window. The explicit equations used for the 13 components of the original feature vector are shown in appendix A. A computer program was written for the Hewlett-Packard 1000 computer to calculate J_1 as a function of displacement in x and y . This computer program was also used to calculate the transformation matrix A . A listing of this program is provided in appendix B. A second computer program was written to calculate the new feature vector Y for the four hundred samples taken from the radar imagery. A listing for this second computer program is given in appendix C.

¹Keinosuke Fukunaga, *Introduction to Statistical Pattern Recognition*, Academic Press, 1972.

RESULTS

In this section some results of numerical calculations are presented. Table 1 shows the results of calculating the value of J_1 for various values of displacement DX and DY for the two classes of forests and cities. The a priori probabilities for the two classes were assumed equal to 0.5.² In table 1 the largest value of J_1 is 29.8525, which is associated with a DX of -3 and a DY of 4. The significance of a maximum value of J_1 occurring at these particular values of displacement is not understood at this time. The values of the displacement DX and DY associated with the largest value of J_1 were utilized to compute a new feature vector Y with two dimensions.

Figure 1 shows a plot of two second-order statistical components for the forest and city samples. The triangles are the results from city calculations; and the x's are the results from forest calculations. This figure clearly shows the necessity for feature selection because the original data for forests and cities is not separated. Figure 2 presents the results after the transformation $Y = AX$ is applied to the original feature vector X. In this figure the x's again represent calculations for forests. We now see that the data from forests is clustered and almost totally separated from the data for cities, represented by the triangles. Even though a few points still overlap, the improvement is very dramatic. The transformation has succeeded in reducing the dimensionality of the original feature vector from 13 to 2 and also in separating the clusters of data for the two classes.

Figure 3 shows the results of the transformation for the two classes of cities and fields. The x's represent the data from cities. Good separation is obtained for the two classes as only a few points are overlapping. Figure 4 shows the results of calculations for the two classes of cities and water. The x's represent the data from cities. In this particular case the two classes are totally separated as well as being clustered fairly well. Figure 5 presents the results for forests and water. The x's represent the data for forests. Again the two classes have clearly been separated by the transformation. Figure 6 shows the results for fields and water, with the data for fields represented by the x's. These two classes have also been totally separated. Figure 7 shows the results for forests and fields with the data for forests represented by the x's. The two classes are not well separated and other features may have to be investigated to obtain better separation. Another possibility would be to try using nonlinear feature selection techniques for these two classes.

²This assumption of equal probability for cities and forests appeared to be appropriate for the area where the samples were taken.

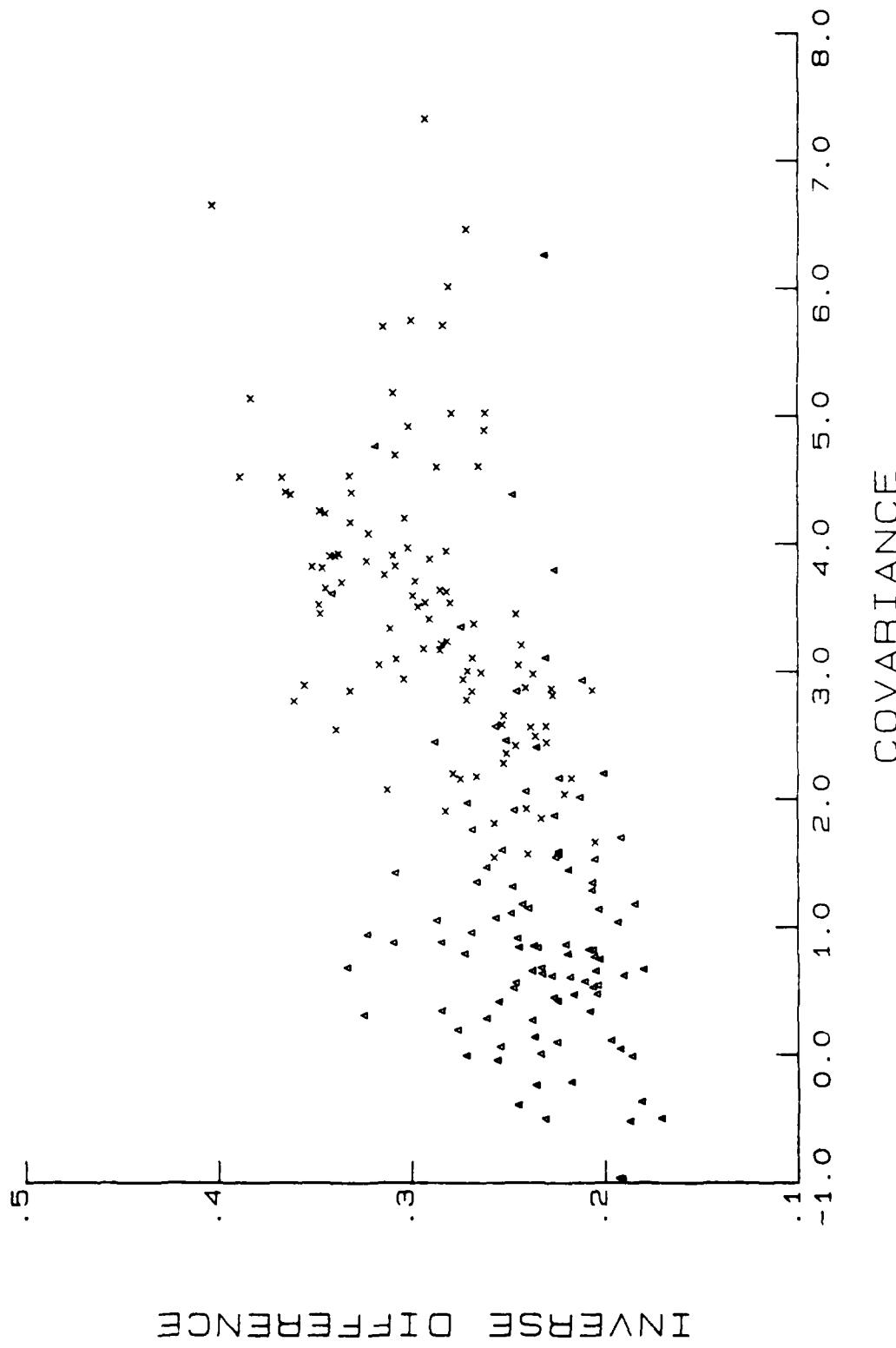
TABLE 1. Values of J_1 for first and second order histogram statistics

DX	DY	J_1
1	0	22.1485
2	0	19.8078
3	0	19.4333
4	0	19.3099
5	0	19.7508
6	0	20.3677
7	0	19.8529
0	1	21.0314
0	2	21.7613
0	3	21.3524
0	4	21.7343
0	5	22.3144
0	6	23.1462
0	7	23.5903
1	1	20.6771
1	2	20.4869
2	2	24.3979
3	2	23.9334
4	2	21.5577
3	3	25.7921
- 1	1	20.2745
- 2	2	24.4373
- 3	2	25.0272
- 3	4	29.8525
4	4	22.7131

CONCLUSIONS

1. The feature selection technique discussed in this report appears to be a powerful tool for the application of pattern recognition to high resolution synthetic aperture radar imagery.
2. In order to separate forests and agricultural fields it appears that the first and second order histogram statistics combined with a linear feature selection technique may not be sufficient as a feature vector.

FIGURE 1. Forest and city samples $DX = -3$ $DY = 4$.



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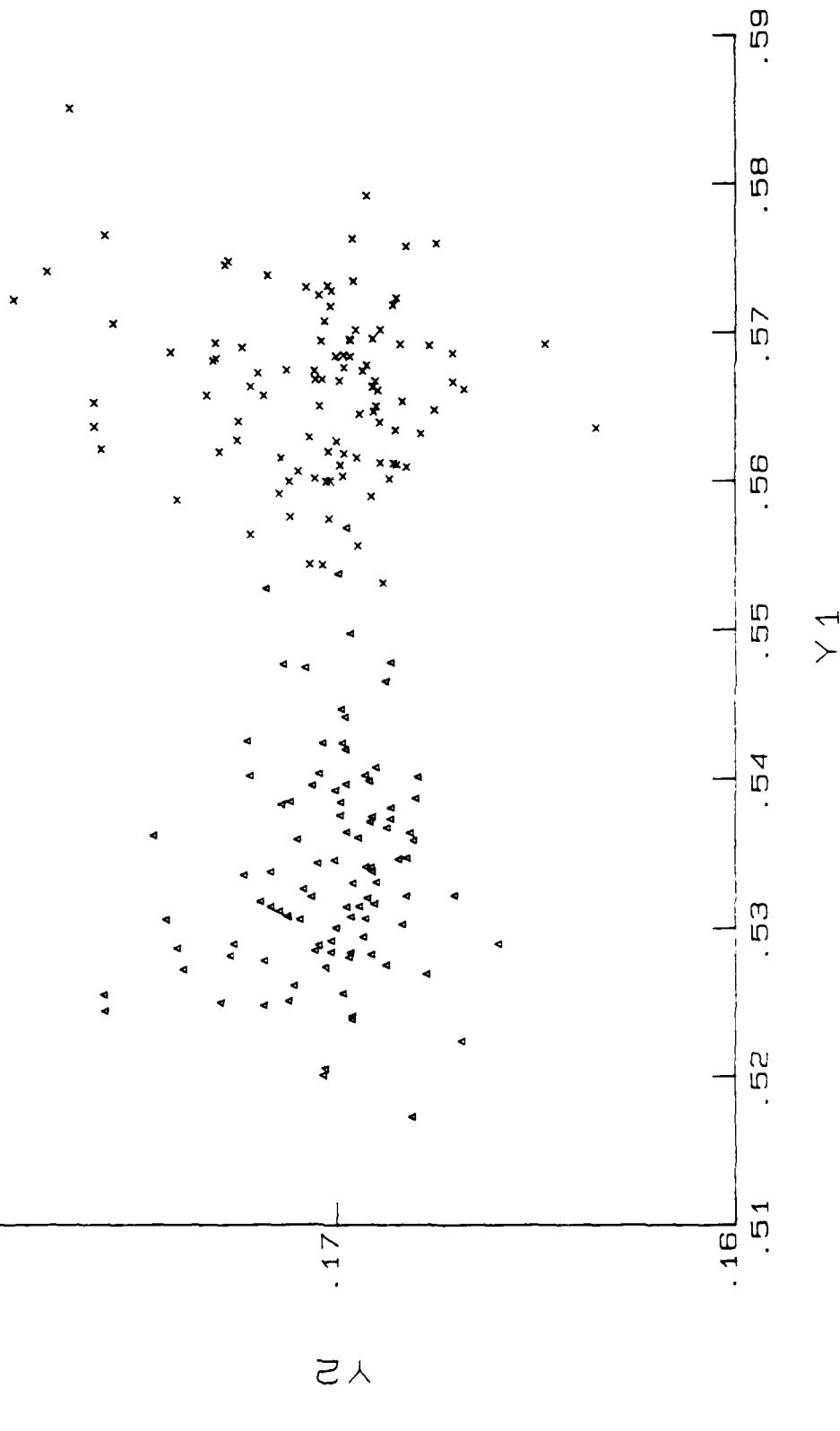


FIGURE 2. Forests and cities $DX = -3$ $DY = 4.$

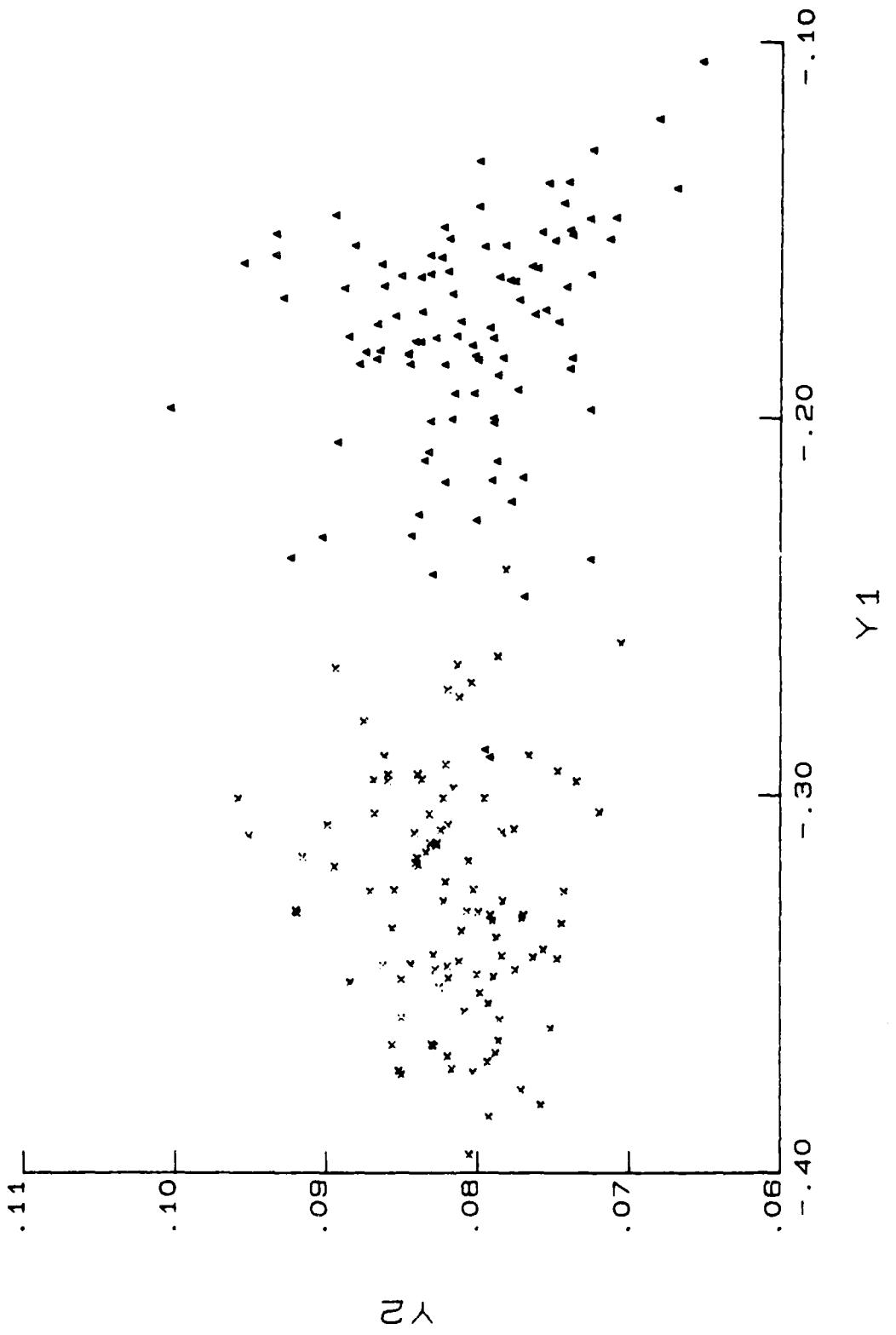
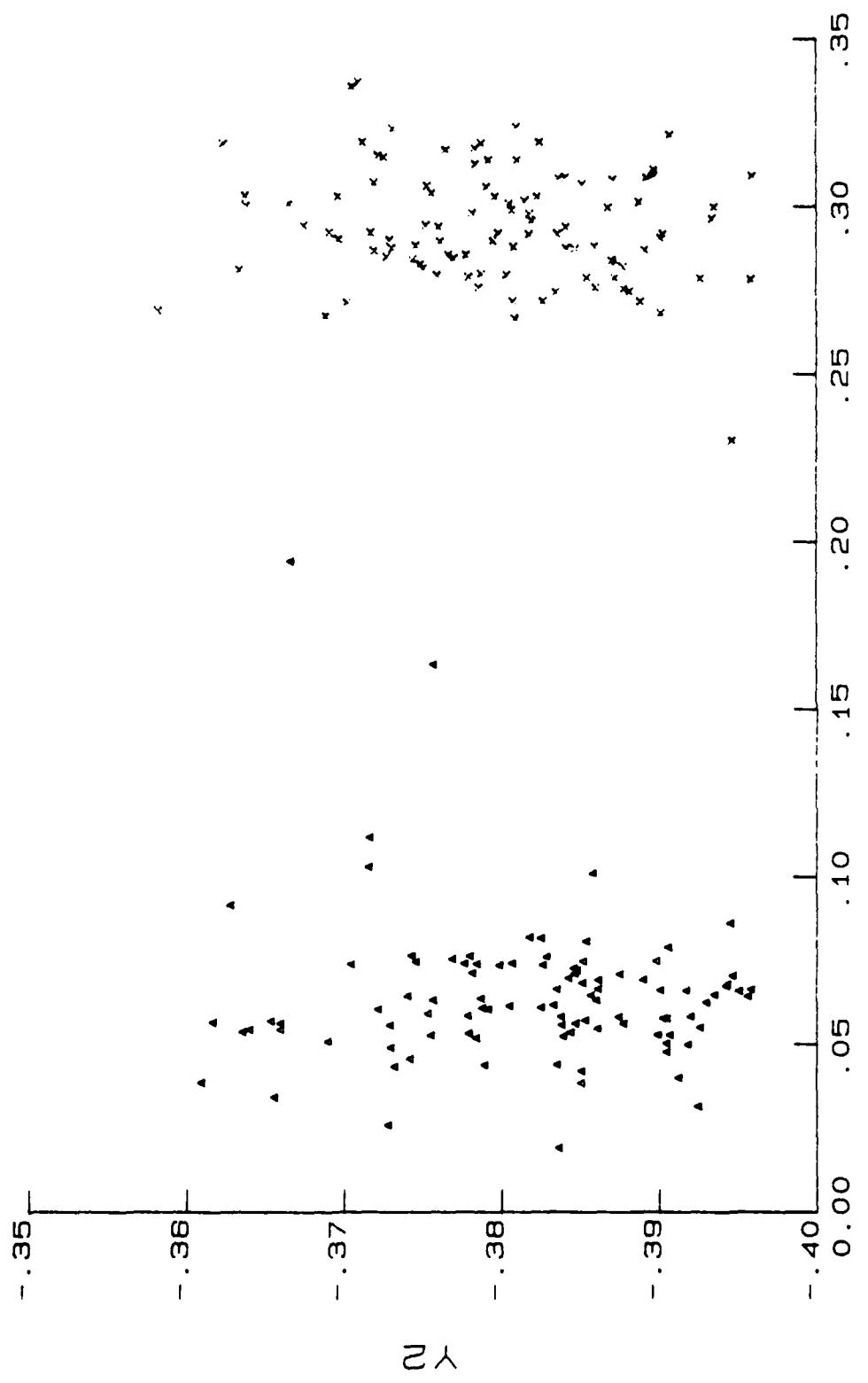


FIGURE 3. Cities and fields $DX = 2$ $DY = 4$.



13

FIGURE 4. Cities and water $DX = 1$ $DY = 0$.

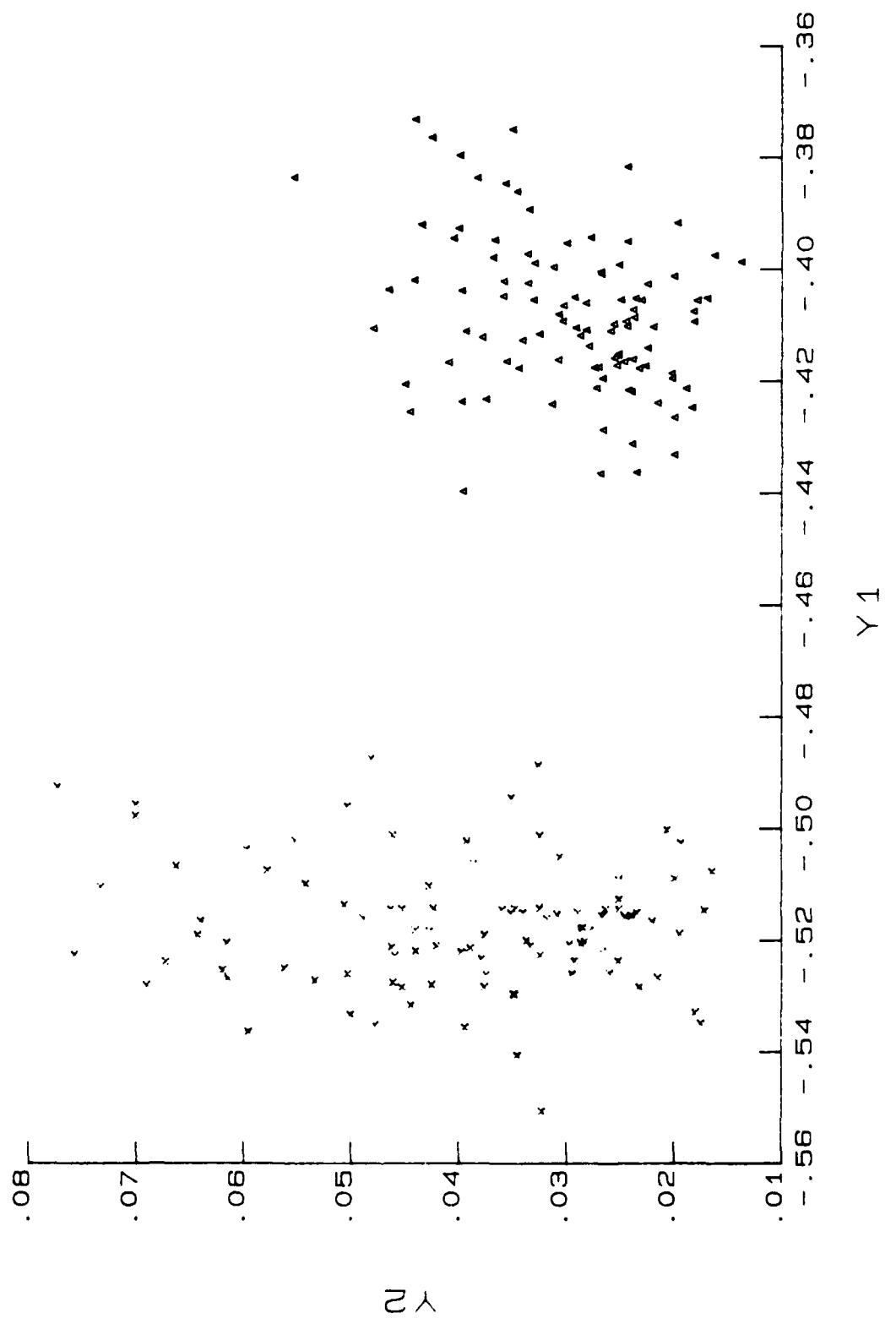


FIGURE 5. Forest and water $DX = 1$ $DY = 0$.

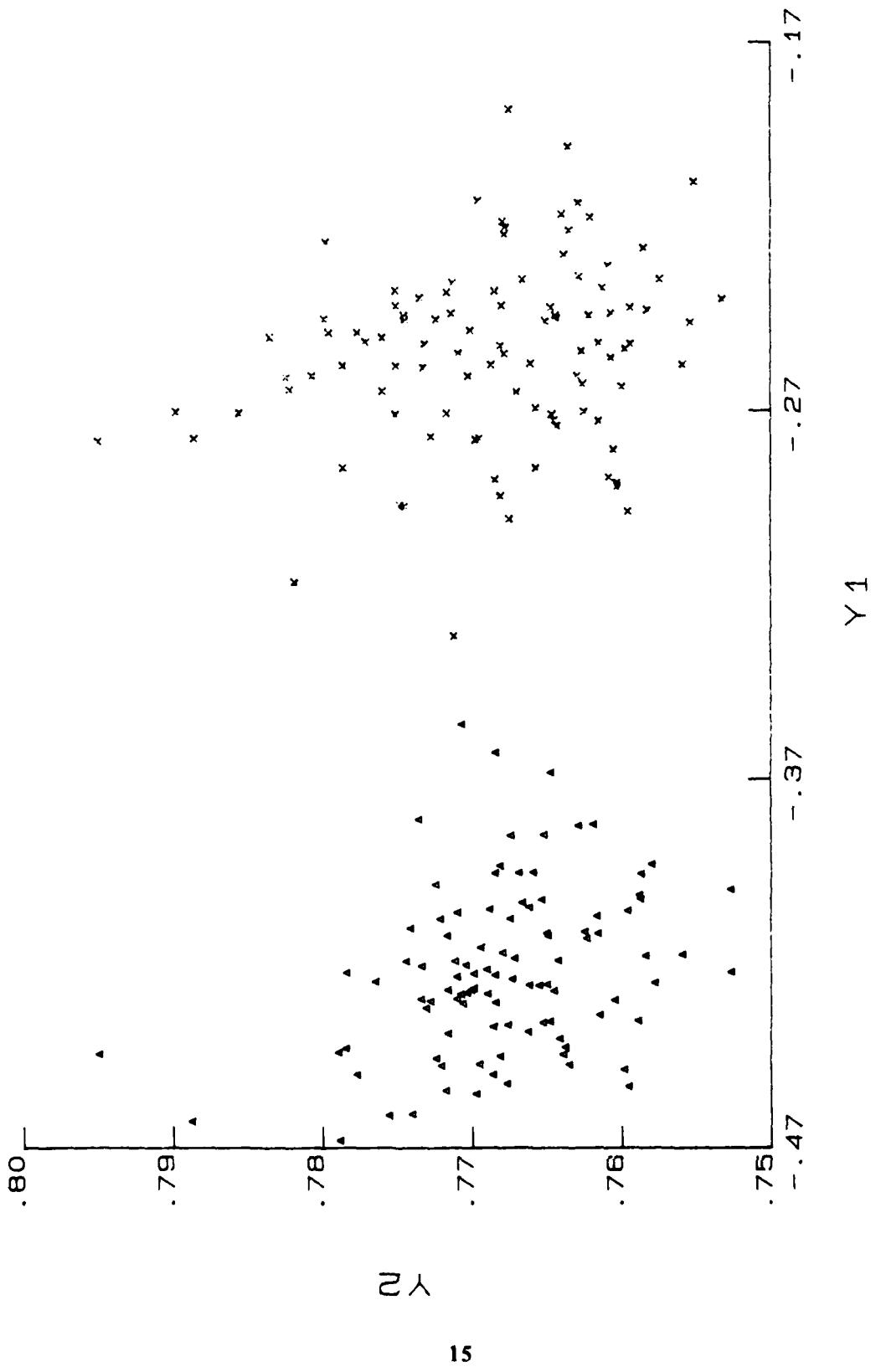


FIGURE 6. Fields and water $DX = 1$ $DY = 0$.

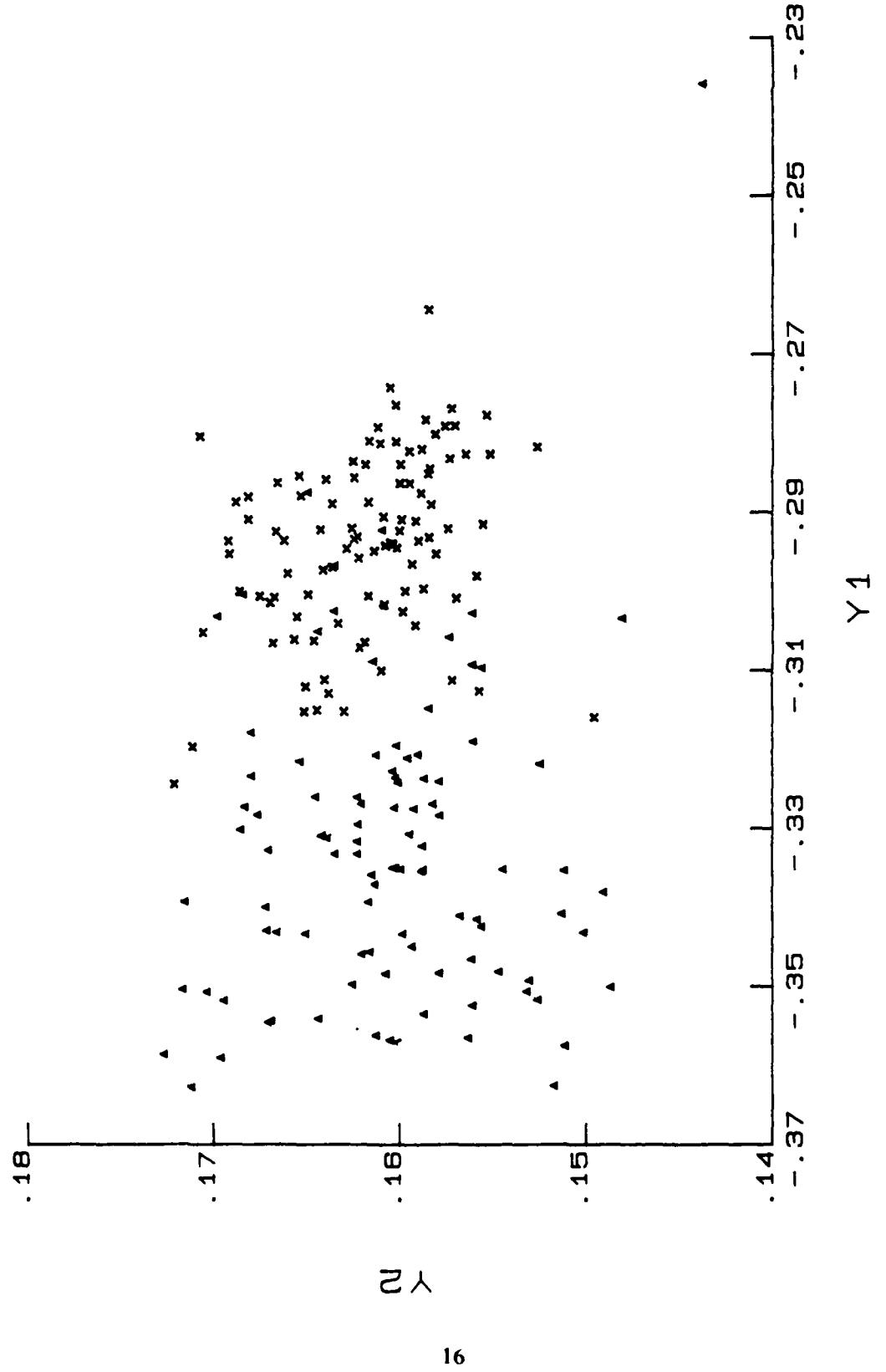


FIGURE 7. Forests and fields $DX = 2$ $DY = 0$.

APPENDIX A

FEATURE VECTOR COMPONENTS

The following first and second order histogram measures were used to construct a thirteen dimensional feature vector:

$$\text{Mean} \quad \bar{b} = \sum_{b=0}^{L-1} bP(b) = x_1$$

$$\text{Variance} \quad \sigma_b^2 = \sum_{b=0}^{L-1} (b - \bar{b})^2 P(b) = x_2$$

$$\text{Skewness} \quad b_S = \frac{1}{\sigma_b^3} \sum_{b=0}^{L-1} (b - \bar{b})^3 P(b) = x_3$$

$$\text{Kurtosis} \quad b_K = \frac{1}{\sigma_b^4} \sum_{b=0}^{L-1} (b - \bar{b})^4 P(b) - 3 = x_4$$

$$\text{Energy} \quad b_N = \sum_{b=0}^{L-1} [P(b)]^2 = x_5$$

$$\text{Entropy} \quad b_E = - \sum_{b=0}^{L-1} P(b) \log_2 [P(b)] = x_6$$

$$\text{Autocorrelation} \quad B_A = \sum_{a=0}^{L-1} \sum_{b=0}^{L-1} abP(a,b) = x_7$$

Covariance $B_C = \sum_{a=0}^{L-1} \sum_{b=0}^{L-1} (a - \bar{a})(b - \bar{b}) P(a,b) = x_8$

Inertia $B_I = \sum_{a=0}^{L-1} \sum_{b=0}^{L-1} (a - b)^2 P(a,b) = x_9$

Absolute Value $B_V = \sum_{a=0}^{L-1} \sum_{b=0}^{L-1} |a - b| P(a,b) = x_{10}$

Inverse Difference $B_D = \sum_{a=0}^{L-1} \sum_{b=0}^{L-1} \frac{P(a,b)}{1 + (a-b)^2} = x_{11}$

Energy $B_N = \sum_{a=0}^{L-1} \sum_{b=0}^{L-1} [P(a,b)]^2 = x_{12}$

Entropy $B_E = -\sum_{a=0}^{L-1} \sum_{b=0}^{L-1} P(a,b) \log_2 [P(a,b)] = x_{13}$

where L is the number of grey levels and $P(b)$ and $P(a,b)$ are given below

$$P(b) = \frac{Q(b)}{M}$$

M is the total number of pixels in the sample window. In this case M was equal to 1024. $Q(b)$ is the number of pixels of greytone b which occur in the sample window.

$$P(a,b) = \frac{Q(a,b)}{M}$$

$Q(a,b)$ is the number of times greytone a is located next to greytone b by the displacement Δx and Δy .

APPENDIX B.

Computer Program for Calculating J_1 and the Transformation Matrix A

```

      C THIS PROGRAM IS FOR CALCULATING J1 AND THE MATRIX A
      C THIS PROGRAM PERFORMS FEATURE EXTRACTION
      C AND FEATURE DISTRIBUTION BY PROCESSING THE IMAGE
      C J1
      C PROGRAM SEECS11000
      DIMENSION D(0:63) IMAGE(0:24), X(0:13), Y(0:14), Z(0:13)
      DIMENSION A(13,2), C(13,13), S(13,13), T(13,13), P(13)
      CALL RMPAR(UNIT)
      CALL RUMAR
      CALL ERUDEC
      CALL ERUDC
      1 READ
      DO 80 NJ=1, N
      WRITE(CLU, 6)
      6 FORMAT("ENTER THE NUMBER OF IMAGE NAME(S) TO BE ANALYZED: ")
      80 READ(CLU, 8) NDATA
      9 FORMAT(3D)
      WRITE(CLU, 9)
      10 FORMAT("ENTER THE FILE NAME FOR THE DATA: ")
      READ(CLU, 10) FILE
      12 FORMAT(3A)
      WRITE(CLU, 12)
      15 FORMAT("DISK LU NUMBER: ")
      READ(CLU, 15) LU
      WRITE(CLU, 15)
      2100 FORMAT("ENTER A VALUE FOR IDX: ")
      READ(CLU, 2100) IDX
      2101 FORMAT(3A)
      WRITE(CLU, 2101)
      2200 FORMAT("ENTER A VALUE FOR IDY: ")
      READ(CLU, 2200) IDY
      WRITE(C6, 1400) IDX, IDY
      1400 FORMAT(2X, "IDX = ", 3A, "IDY = ", 3A)
      WRITE(C6, 1400) LU, FILE
      2100 FORMAT(3X, 3A)
      2101 FORMAT(3A)
      CALL OPEN(CLU, TERR, FILE, 0, 0, CNUC)
      IF(TERR.LT.0) GO TO 2000
      GO TO 14
      2000 WRITE(CLU, 2000) TERR
      2100 FORMAT("OPEN FILE ERROR", I5)
      GO TO 999
      14 ICONT=1
      15 J=1
      16 DO 16 J=1, 8
      17 CALL READ(CLU, TERR, IMAGE(0:3))
      18 J=J+128
      IF(TERR.LT.0) GO TO 2000
      GO TO 18
      2000 ICONT=NDATA
      WRITE(CLU, 2020) TERR
      2020 FORMAT("READ FILE ERROR", I5)
      GO TO 999
      18 INGRAY=16
      CALL JSCL(IMAGE, IMAGE(1:0:24), 0, 15)
      CALL FEVAL(IMAGE, AVG, VAR, SRW, XKT, ENGL, ALL, COV, XII,
      TABSS, XID, ENR, INGRAY, IDX, IDY)

```

APPENDIX B (Continued)

```

0059      C
0060      C   SUBROUTINE TO READ IN AN ARRAYS OF COEFFICIENTS AND SCALES
0061      C
0062      XCOUNT(1)=N
0063      XCOUNT(2)=VAR
0064      XCOUNT(3)=SKW
0065      XCOUNT(4)=KKT
0066      XCOUNT(5)=SIGMA
0067      XCOUNT(6)=ENT
0068      XCOUNT(7)=HLL
0069      XCOUNT(8)=LIV
0070      XCOUNT(9)=KLE
0071      XCOUNT(10)=BRS
0072      XCOUNT(11)=XED
0073      XCOUNT(12)=ENTY
0074      XCOUNT(13)=LNR
0075      IF(COUNT(NDATA)>0) GOTO 20
0076      COUNT=COUNT+1
0077      GO TO 15
0078      C2  CALL CLONE(C1D0,0)
0079      WRITE(C1,24)NC1,TEEE
0080      24 FORMAT("N=" I1.3A 2E-3)
0081      CALL COVERNDATA(I, P, Q, N, X, D5)
0082      WRITE(6,700)D5
0083      70 FORMAT(1X," THE INTERSET DISTANCE = "ES15.6)
0084      WRITE(6,140)
0085      140 FORMAT(1X,"COVARIANCE MATRIX")
0086      DO 142 K=1,3
0087      WRITE(6,144)C(K,M,N), M=1,3
0088      144 FORMAT(1X,(3E8.4),1X)
0089      145 CONTINUE
0090      80 CONTINUE
0091      150 DO 152 NJ=1, NC
0092      WRITE(C1,154)NJ
0093      154 FORMAT("NJ=" I1.2X "ENTER THE APRIORI PROBABILITY OF THIS CLASS")
0094      READ(C1,180)P(NJ)
0095      180 FORMAT(E18.8)
0096      WRITE(6,156)NT,P(NJ)
0097      156 FORMAT(1X,"NJ=" I1.2X "PN(J) = " E10.8)
0098      152 CONTINUE
0099      DO 90 J=1,NC-1
0100      DO 90 I=J+1,NC
0101      90 S2(I,J)=0.0
0102      DO 72 J=1,NC-1
0103      DO 72 I=J+1,NC
0104      DO 72 M=1,3S
0105      S2(I,J)=S2(I,J)+C(M,I)*C(M,J)
0106      72 CONTINUE
0107      DO 93 I=1, NC-1
0108      DO 93 J=I+1,NC
0109      S2(I,J)=SQRT(S2(I,J))
0110      93 CONTINUE
0111      WRITE(6,73)
0112      73 FORMAT(1X,"THE INTERSET DISTANCES")
0113      DO 92 I=1,NC-1
0114      DO 92 J=I+1,NC
0115      WRITE(6,74)I,J,S2(I,J)
0116      74 FORMAT(1X,"I=" I1.2X,"J=" I1.2X,"D(I,J) = " E15.8)
0117      92 CONTINUE
0118      DO 30 I=1,13

```

APPENDIX B (Continued)

```

0119      DO 30 J=1,13
0120      S2(1,J)=0.0
0121      30 S1(1,J)=0.0
0122      DO 32 I=1,13
0123      DO 32 J=1,13
0124      DO 32 K=1,NC
0125      32 S2(I,J)=S2(I,J)+E(I,J,K)*P(K)
0126      DO 34 J=1,13
0127      34 XH0(J)=0.0
0128      DO 40 J=1,13
0129      DO 40 M=1,13
0130      40 S1(J,M)=C(J,M,1)+C(J,M,2)+(A(J,1)-A(J,2))*(A(M,1)-A(M,2))
0131      DO 160 J=1,13
0132      DO 160 I=1,13
0133      K=I+13*(J-1)
0134      R(K)=S1(I,J)
0135      E(K)=S2(I,J)
0136      160 CONTINUE
0137      DO 41 I=1,13
0138      41 XM0(I)=0.0
0139      CALL NRROUT(13,R,B,XM0,I)
0140      DO 165 J=1,13
0141      DO 165 I=1,13
0142      K=I+13*(J-1)
0143      S2(I,J)=E(K)
0144      165 CONTINUE
0145      WRITE(6,42)
0146      42 FORMAT(2X,"EIGENVALUES")
0147      WRITE(6,46)(XM0(I),I=1,13)
0148      46 FORMAT(1X,13(F9.3,1X))
0149      WRITE(6,48)
0150      48 FORMAT(2X,"EIGENVECTORS")
0151      DO 50 I=1,13
0152      WRITE(6,52)(S2(I,J),J=1,13)
0153      52 FORMAT(1X,13(F8.5,1X))
0154      50 CONTINUE
0155      64 WRITE(LU,52)
0156      52 FORMAT("ENTER A VALUE FOR M<E,N<")
0157      READ(LU,43)
0158      XJ1=0.0
0159      DO 54 I=1,M
0160      54 XJ1=XJ1+XM0(I)
0161      WRITE(6,56)XJ1,M
0162      56 FORMAT(2X,"THE VALUE OF J1=",F10.4,5X,"M=",I1)
0163      CALL IRMAT(S2,S1,13,13,0)
0164      WRITE(6,58)
0165      58 FORMAT(2X,"THE TRANSFORMATION MATRIX A")
0166      DO 60 I=1,M
0167      WRITE(6,46)(S1(I,J),J=1,13)
0168      60 CONTINUE
0169      WRITE(LU,62)
0170      62 FORMAT("DO YOU WANT TO CHANGE THE VALUE OF M? IF YES TYPE 1 IF
0171      3 NO TYPE 2")
0172      READ(LU,47)IZ
0173      47 FORMAT(I1)
0174      IF (IZ.EQ.1) GO TO 64
0175      WRITE(LU,62)
0176      62 FORMAT("DO YOU WANT TO CHANGE THE VALUES OF THE APRIORI PROBABIL
0177      ITIES? IF YES TYPE 1 IF NO TYPE 2")
0178      READ(LU,47)IAZ

```

APPENDIX B (Continued)

```

0179      TECJAZ,F0,1)GO TO 150
0180      WRITE(0,68)
0181      68 FORMAT("DO YOU WANT TO CHANGE THE VALUES OF TDX AND TDY? IF YES
0182      TYPE 1, IF NO TYPE 2.")
0183      READ(0,4)X2
0184      TECJXZ,EN,1)GO TO 1
0185      999 STOP
0186      END
0187      SUBROUTINE COVCRK(LDU,AU,N,NL,X,D)
0188      DIMENSION X(100,13),U(13,13),AU(13,13),COVCR(13,20),AVC(3
0189      1,2),VAR(13)
0190      KK=KK
0191      DO 50 K=1,N
0192      ACK=0.0
0193      DO 45 J=1,KK
0194      45 ACK=ACK+X(J,K)
0195      50 ACK=ACK/ACK
0196      DO 55 K=1,N
0197      VAR(K)=0.0
0198      DO 54 J=1,KK
0199      54 VAR(K)=VAR(K)+X(J,K)
0200      55 CONTINUE
0201      DO 60 K=1,N
0202      60 VAR(K)=VAR(K)/ACK
0203      D=0.0
0204      DO 65 K=1,N
0205      65 D=D+VAR(K)
0206      D=2.0*D
0207      DO 70 K=1,N
0208      70 D=1.05*D
0209      D=(D-ACK)/ACK
0210      110 CONTINUE
0211      DO 120 K=1,N
0212      120 D25=M-1.0
0213      LCK=0.0
0214      DO 130 I=1,KK
0215      130 LCK=LCK+ACK*I
0216      LCK=M-LCK
0217      140 CONTINUE
0218      150 CONTINUE
0219      DO 155 K=1,N
0220      155 M=1.0
0221      LCK=N-LCK
0222      155 LCK=M-N-LCK
0223      END
0224      SUBROUTINE NREDUCE(K,L,NL,K2)
0225      SUBROUTINE OCTAVE(NL,K,L,NL,K2)
0226      K=0
0227      DO 160 I=1,NL
0228      L=MAX(I,L)
0229      DO 170 J=I+1,NL
0230      L=L+1
0231      K=K+1
0232      END
0233      DO 180 R=L-K+1,L
0234      180 R=R+1
0235      CALL ELEVENTH(R,L)
0236      C=0
0237      DO 190 I=1,L
0238      190 C=C+1

```

APPENDIX B
(Continued)

0.239	110 XE(1D+1)0.05R(EA(1C))
0.240	K=0
0.241	DO 115 J=1,M
0.242	DO 115 L=1,M
0.243	K=K+1
0.244	115 R(K)=X(K)*X(L)
0.245	DO 120 I=1,M
0.246	N=N+1
0.247	DO 120 J=1,N
0.248	N=M*(L-1)
0.249	C=M*(L-1)+1
0.250	X(C)=0.0
0.251	DO 120 K=1,M
0.252	N=0.0+1
0.253	M=N+1
0.254	120 X(C)=X(C)+R(K)*X(C),I
0.255	I=0
0.256	DO 130 J=1,I
0.257	DO 130 L=1,J
0.258	N=L=0
0.259	N=M*(J-1)
0.260	L=L+1
0.261	DO 130 K=1,M
0.262	M=N+1
0.263	N=M+1
0.264	130 N=N+1
0.265	130 C=C+X(K)*X(L)
0.266	130 L=L+1
0.267	C=0.0
0.268	DO 140 I=1,M
0.269	I=I+1
0.270	140 X(C)=X(C)
0.271	DO 150 J=1,I
0.272	N=L=0
0.273	N=M*(J-1)
0.274	L=L+1
0.275	DO 150 K=1,M
0.276	M=N+1
0.277	N=M+1
0.278	150 N=N+1
0.279	150 C=C+X(K)*X(L)
0.280	C=0.0
0.281	DO 160 I=1,M
0.282	I=I+1
0.283	DO 160 J=1,I
0.284	N=L=0
0.285	DO 160 K=1,M
0.286	M=N+1
0.287	N=M+1
0.288	160 N=N+1
0.289	160 C=C+X(K)*X(L)
0.290	C=0.0
0.291	DO 170 I=1,M
0.292	I=I+1
0.293	DO 170 J=1,I
0.294	N=L=0
0.295	DO 170 K=1,M
0.296	M=N+1
0.297	N=M+1
0.298	170 N=N+1
0.299	170 C=C+X(K)*X(L)
0.300	C=0.0
0.301	DO 180 I=1,M
0.302	I=I+1
0.303	DO 180 J=1,I
0.304	N=L=0
0.305	DO 180 K=1,M
0.306	M=N+1
0.307	N=M+1
0.308	180 N=N+1
0.309	180 C=C+X(K)*X(L)
0.310	C=0.0
0.311	DO 190 I=1,M
0.312	I=I+1
0.313	DO 190 J=1,I
0.314	N=L=0
0.315	DO 190 K=1,M
0.316	M=N+1
0.317	N=M+1
0.318	190 N=N+1
0.319	190 C=C+X(K)*X(L)
0.320	C=0.0
0.321	DO 200 I=1,M
0.322	I=I+1
0.323	DO 200 J=1,I
0.324	N=L=0
0.325	DO 200 K=1,M
0.326	M=N+1
0.327	N=M+1
0.328	200 N=N+1
0.329	200 C=C+X(K)*X(L)
0.330	C=0.0
0.331	DO 210 I=1,M
0.332	I=I+1
0.333	DO 210 J=1,I
0.334	N=L=0
0.335	DO 210 K=1,M
0.336	M=N+1
0.337	N=M+1
0.338	210 N=N+1
0.339	210 C=C+X(K)*X(L)
0.340	C=0.0
0.341	DO 220 I=1,M
0.342	I=I+1
0.343	DO 220 J=1,I
0.344	N=L=0
0.345	DO 220 K=1,M
0.346	M=N+1
0.347	N=M+1
0.348	220 N=N+1
0.349	220 C=C+X(K)*X(L)
0.350	C=0.0
0.351	DO 230 I=1,M
0.352	I=I+1
0.353	DO 230 J=1,I
0.354	N=L=0
0.355	DO 230 K=1,M
0.356	M=N+1
0.357	N=M+1
0.358	230 N=N+1
0.359	230 C=C+X(K)*X(L)
0.360	C=0.0
0.361	DO 240 I=1,M
0.362	I=I+1
0.363	DO 240 J=1,I
0.364	N=L=0
0.365	DO 240 K=1,M
0.366	M=N+1
0.367	N=M+1
0.368	240 N=N+1
0.369	240 C=C+X(K)*X(L)
0.370	C=0.0
0.371	DO 250 I=1,M
0.372	I=I+1
0.373	DO 250 J=1,I
0.374	N=L=0
0.375	DO 250 K=1,M
0.376	M=N+1
0.377	N=M+1
0.378	250 N=N+1
0.379	250 C=C+X(K)*X(L)
0.380	C=0.0
0.381	DO 260 I=1,M
0.382	I=I+1
0.383	DO 260 J=1,I
0.384	N=L=0
0.385	DO 260 K=1,M
0.386	M=N+1
0.387	N=M+1
0.388	260 N=N+1
0.389	260 C=C+X(K)*X(L)
0.390	C=0.0
0.391	DO 270 I=1,M
0.392	I=I+1
0.393	DO 270 J=1,I
0.394	N=L=0
0.395	DO 270 K=1,M
0.396	M=N+1
0.397	N=M+1
0.398	270 N=N+1
0.399	270 C=C+X(K)*X(L)
0.400	C=0.0
0.401	DO 280 I=1,M
0.402	I=I+1
0.403	DO 280 J=1,I
0.404	N=L=0
0.405	DO 280 K=1,M
0.406	M=N+1
0.407	N=M+1
0.408	280 N=N+1
0.409	280 C=C+X(K)*X(L)
0.410	C=0.0
0.411	DO 290 I=1,M
0.412	I=I+1
0.413	DO 290 J=1,I
0.414	N=L=0
0.415	DO 290 K=1,M
0.416	M=N+1
0.417	N=M+1
0.418	290 N=N+1
0.419	290 C=C+X(K)*X(L)
0.420	C=0.0
0.421	DO 300 I=1,M
0.422	I=I+1
0.423	DO 300 J=1,I
0.424	N=L=0
0.425	DO 300 K=1,M
0.426	M=N+1
0.427	N=M+1
0.428	300 N=N+1
0.429	300 C=C+X(K)*X(L)
0.430	C=0.0
0.431	DO 310 I=1,M
0.432	I=I+1
0.433	DO 310 J=1,I
0.434	N=L=0
0.435	DO 310 K=1,M
0.436	M=N+1
0.437	N=M+1
0.438	310 N=N+1
0.439	310 C=C+X(K)*X(L)
0.440	C=0.0
0.441	DO 320 I=1,M
0.442	I=I+1
0.443	DO 320 J=1,I
0.444	N=L=0
0.445	DO 320 K=1,M
0.446	M=N+1
0.447	N=M+1
0.448	320 N=N+1
0.449	320 C=C+X(K)*X(L)
0.450	C=0.0
0.451	DO 330 I=1,M
0.452	I=I+1
0.453	DO 330 J=1,I
0.454	N=L=0
0.455	DO 330 K=1,M
0.456	M=N+1
0.457	N=M+1
0.458	330 N=N+1
0.459	330 C=C+X(K)*X(L)
0.460	C=0.0
0.461	DO 340 I=1,M
0.462	I=I+1
0.463	DO 340 J=1,I
0.464	N=L=0
0.465	DO 340 K=1,M
0.466	M=N+1
0.467	N=M+1
0.468	340 N=N+1
0.469	340 C=C+X(K)*X(L)
0.470	C=0.0
0.471	DO 350 I=1,M
0.472	I=I+1
0.473	DO 350 J=1,I
0.474	N=L=0
0.475	DO 350 K=1,M
0.476	M=N+1
0.477	N=M+1
0.478	350 N=N+1
0.479	350 C=C+X(K)*X(L)
0.480	C=0.0
0.481	DO 360 I=1,M
0.482	I=I+1
0.483	DO 360 J=1,I
0.484	N=L=0
0.485	DO 360 K=1,M
0.486	M=N+1
0.487	N=M+1
0.488	360 N=N+1
0.489	360 C=C+X(K)*X(L)
0.490	C=0.0
0.491	DO 370 I=1,M
0.492	I=I+1
0.493	DO 370 J=1,I
0.494	N=L=0
0.495	DO 370 K=1,M
0.496	M=N+1
0.497	N=M+1
0.498	370 N=N+1
0.499	370 C=C+X(K)*X(L)
0.500	C=0.0

APPENDIX B
(Continued)

```

0299      VAR=0,0
0300      SKW=0,0
0301      XK1=0,0
0302      FNG=0,0
0303      FNT=0,0
0304      AUT=0,0
0305      COV=0,0
0306      XTE=0,0
0307      ABS5=0,0
0308      X1D=0,0
0309      FNY=0,0
0310      ENR=0,0
0311      DO 30 J=1,32
0312      DO 30 I=1,32
0313      M=1,1
0314      30 TARC(I,J)=IMAGE(I+M*32)
0315      M=3024
0316      DO 20 K=1,NGRAY
0317      20 CONT(K)=0,0
0318      DO 30 I=1,32
0319      DO 30 J=1,32
0320      TR=TARC(I,J)
0321      CONTCR+1=CONT(CTR+I+NJ+1)
0322      30 CONTINUE
0323      XM=0
0324      DO 32 K=1,NGRAY
0325      DO 32 M=1,NGRAY
0326      32 CONT(K,M)=0,0
0327      MX=32+1DX
0328      1F (32-MX)12,14,16
0329      12 NXCR=MX+32
0330      NXB=1
0331      NXF=32-NXCR
0332      GO TO 19
0333      14 NXH=1
0334      NXL=32
0335      GO TO 19
0336      16 NXR=32-NX
0337      NXB=1+NXR
0338      NXL=33-NXH
0339      18 NFT=32+1DX
0340      1F (32-NY)12,14,16
0341      21 NYUR=NY-32
0342      NYB=1
0343      NYF=32-NYUR
0344      NYL=33-NYH
0345      23 NYR=1
0346      NYL=32
0347      NYF=32-NY
0348      NYB=1+NYUR
0349      NYL=33-NYH
0350      25 NYURENDF
0351      NYL=32-NY
0352      NYF=32-NYUR
0353      NYB=1+NYUR
0354      NYL=33-NYH
0355      27 NYURENDF
0356      NYL=32-NY
0357      NYF=32-NYUR
0358      NYB=1+NYUR
0359      NYL=33-NYH
0360      29 NYURENDF
0361      NYL=32-NY
0362      NYF=32-NYUR

```

APPENDIX B
(Continued)

```

0359      40 P(K)=CONT(K)/XM
0360      DO 45 K=1,NGRAY
0361      45 AVG=AVG+(K-1)*P(K)
0362      DO 50 K=1,NGRAY
0363      50 VAR=VAR+((K-1-AVG)**2)*P(K)
0364      STD=SQR1(VAR)
0365      DO 55 K=1,NGRAY
0366      55 SKW=SKW+((K-1-AVG)**3)*P(K)
0367      SKW=SKW/STD**3
0368      DO 60 K=1,NGRAY
0369      60 XKT=XKT+((K-1-AVG)**4)*P(K)
0370      XK1=(XKT/(STD**4))-3.
0371      DO 65 K=1,NGRAY
0372      65 ENG=ENG+P(K)*P(K)
0373      DO 70 K=1,NGRAY
0374      IF(P(K).EQ.0.0)GO TO 70
0375      ENT=ENT+3.321929*P(K)*ALOGT(P(K))
0376      /0 CONTINUE
0377      ENT=FNT
0378      DO 75 K=1,NGRAY
0379      DO 75 M=1,NGRAY
0380      75 PP(K,M)=CONT(K,M)/XM
0381      DO 80 K=1,NGRAY
0382      DO 80 M=1,NGRAY
0383      AUT=AUT+(K-1)*(M-1)*PP(K,M)
0384      80 CONTINUE
0385      AVGK=0.0
0386      AVGM=0.0
0387      DO 85 K=1,NGRAY
0388      DO 85 M=1,NGRAY
0389      AVGK=AVGK+(K-1)*PP(K,M)
0390      85 AVGM=AVGM+(M-1)*PP(K,M)
0391      DO 90 K=1,NGRAY
0392      DO 90 M=1,NGRAY
0393      90 XIE=XIE+(K-M)*(K-M)*PP(K,M)
0394      DO 95 K=1,NGRAY
0395      DO 95 M=1,NGRAY
0396      ARSS=ARSS+(JABS(K-M))*PP(K,M)
0397      95 CONTINUE
0398      DO 100 K=1,NGRAY
0399      DO 100 M=1,NGRAY
0400      100 XJD=XJD+PP(K,M)/(1.0+(K-M)**2)
0401      DO 105 K=1,NGRAY
0402      DO 105 M=1,NGRAY
0403      105 ENY=ENY+PP(K,M)*PP(K,M)
0404      DO 110 K=1,NGRAY
0405      DO 110 M=1,NGRAY
0406      IF(PP(K,M).EQ.0.0)GO TO 110
0407      ENR=ENR+3.321929*PP(K,M)*ALOGT(PP(K,M))
0408      110 CONTINUE
0409      ENR=-ENR
0410      DO 115 K=1,NGRAY
0411      DO 115 M=1,NGRAY
0412      COV=COV+(K-1-AVGK)*(M-1-AVGM)*PP(K,M)
0413      115 CONTINUE
0414      RETURN
0415      END
0416      END$
0417

```

APPENDIX C.

Computer Program for Calculating the Transformed Feature Vector $Y = AX$

```

A YAXIR T=00004 IS ON CRU0011 USING 00013 BLKS R=0000

0001  FIN4.L
0002  CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC
0003  C THIS PROGRAM COMPUTES THE TRANSFORMATION          C
0004  C Y=AX WHERE A IS COMPUTED FROM ANOTHER          C
0005  C PROGRAM AND IS INPUT HERE          C
0006  CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC
0007  PROGRAM YAXIR(3,1000)
0008  DIMENSION LU00T(5),IMAGE(1024),X(100,13),Y(100,2),A(2,13)
0009  DIMENSION IDCB(144),FILE(3),AM(2),VAR(2)
0010  DIMENSION AV(6,6),D(2,2)
0011  EQUIVALENCE (IMAGE,Y)
0012  CALL RMPAR(LU00T)
0013  LU=LU00T(1)
0014  CALL ERLU(LU)
0015  NC=2
0016  4 FORMAT(1I1)
0017  WRITE(LU,6)
0018  6 FORMAT("ENTER THE VALUE FOR M THE NUMBER OF ROWS IN THE TRANSFOR
0019  MATION MATRIX A")
0020  READ(LU,4)M
0021  WRITE(LU,8)
0022  8 FORMAT("ENTER THE TRANSFORMATION MATRIX A")
0023  DO 10 I=1,M
0024  DO 10 J=1,13
0025  WRITE(LU,12)I,J
0026  12 FORMAT("I=",12.2X,"J=",12,"A(I,J)=")
0027  READ(LU,14)A(I,J)
0028  14 FORMAT(E9.3)
0029  10 CONTINUE
0030  WRITE(6,79)
0031  79 FORMAT(1X,"THE TRANSFORMATION MATRIX A")
0032  DO 78 I=1,M
0033  WRITE(6,77)(A(I,J),J=1,13)
0034  77 FORMAT(1X,13(E9.3,1X))
0035  78 CONTINUE
0036  DO 80 NJ=1,NC
0037  WRITE(LU,16)
0038  16 FORMAT("ENTER THE NUMBER OF IMAGE SAMPLES TO BE ANALYZED (I.E.,100")
0039  READ(LU,18)NDATA
0040  18 FORMAT(1I3)
0041  WRITE(LU,20)
0042  20 FORMAT("ENTER THE FILE NAME FOR THE DATA SET")
0043  READ(LU,22)FILE
0044  22 FORMAT(3A2)
0045  WRITE(LU,15)
0046  15 FORMAT("DISK LU NUMBER?")
0047  READ(LU,2101)IDLU
0048  WRITE(LU,2100)
0049  2100 FORMAT("ENTER A VALUE FOR IDX")
0050  READ(LU,2101)IDX
0051  2101 FORMAT(1I2)
0052  WRITE(LU,2200)
0053  2200 FORMAT("ENTER A VALUE FOR IDY")
0054  READ(LU,2101)IDY
0055  WRITE(6,1400)IDX,1DY
0056  1400 FORMAT(2X,"IDX=",12.5X,"IDY=",12)
0057  WRITE(6,200)FILE
0058  200 FORMAT(1X,3A2)

```

APPENDIX C.
(Continued)

```

0059      CALL OPEN(1DCH,IERR,1FILE,0,0,-1DLU)
0060      IF(IERR.LT.0)GO TO 2000
0061      GO TO 24
0062 2000  WRITE(LU,2010)IERR
0063 2010  FORMAT("OPEN FILE ERROR",I5)
0064      GO TO 999
0065      24 1CONT=1
0066      CALL LABIN(1DCH,6)
0067      13 J=1
0068      DO 19 I=1,8
0069      CALL READF(1DCH,IERR,1IMAGE(J))
0070      19 J=J+128
0071      IF(IERR.LT.0)GO TO 3000
0072      GO TO 26
0073 3000  ICONT=NDATA
0074      WRITE(LU,3020)IERR
0075 3020  FORMAT("READ FILE ERROR",I5)
0076      GO TO 999
0077      26 NGRAY=16
0078      CALL JSCL(1IMAGE,1IMAGE,1024,0,15)
0079      CALL FEVEC(1IMAGE,AVG,VAR,SKW,XKT,ENG,ENT,AUT,COV,XIE,AHSS,XJD,
0080      LENY,ENR,NGRAY,1DX,1DY)
0081      X(1CONT,1)=AVG
0082      X(1CONT,2)=VAR
0083      X(1CONT,3)=SKW
0084      X(1CONT,4)=XKT
0085      X(1CONT,5)=ENG
0086      X(1CONT,6)=ENT
0087      X(1CONT,7)=AUT
0088      X(1CONT,8)=COV
0089      X(1CONT,9)=XIE
0090      X(1CONT,10)=AHSS
0091      X(1CONT,11)=XJD
0092      X(1CONT,12)=ENY
0093      X(1CONT,13)=ENR
0094      IF(1CONT-NDATA)>28,30,30
0095 30 1CONT=1CONT+1
0096      GO TO 13
0097 30  WRITE(LU,320N1,1FILE
0098 32  FORMAT("NJ=",I1,3X,3A2)
0099      DO 34 J=1,NDATA
0100      DO 34 K=1,M
0101      34 Y(J,K)=0.0
0102      DO 36 I=1,NDATA
0103      DO 36 K=1,M
0104      DO 36 MK=1,13
0105      36 YC(J,K)-YC(I,K)+ACK(MK)*XC(I,MK)
0106      CALL GPLOT(YC(1,1),YC(1,2),100,11,N1)
0107      DO 38 K=1,M
0108      AM(K)=0.0
0109      NK=NDATA
0110      DO 45 J=1,NDATA
0111      45 AM(K)=AM(K)+Y(J,K)
0112      50 AM(K)=AM(K)/NK
0113      DO 55 K=1,M
0114      VAR(K)=0.0
0115      DO 54 I=1,NDATA
0116      54 VAR(K)=VAR(K)+(YC(I,K)-AM(K))**2
0117      55 CONTINUE
0118      DO 60 K=1,M

```

APPENDIX C.
(Continued)

```
0119      60 VAR(K)=VAR(K)/(XK-1.)  
0120      D3=0.0  
0121      DO 65 K=1,M  
0122      65 D3=D3+VAR(K)  
0123      D3=2.*D3  
0124      DO 70 K=1,M  
0125      70 AV(K,NJ)=AM(K)  
0126      80 CONTINUE  
0127      CALL GPLUT(Y(1,1),Y(1,2),-100,-LU,2)  
0128      DO 81 I=1,NC-1  
0129      DO 81 J=I+1,NC  
0130      81 D(I,J)=0.0  
0131      DO 82 I=1,NC-1  
0132      DO 82 J=I+1,NC  
0133      DO 82 K=1,M  
0134      82 D(I,J)=D(I,J)+(AV(K,I)-AV(K,J))**2  
0135      DO 83 I=1,NC-1  
0136      DO 83 J=I+1,NC  
0137      83 D(I,J)=SQRT(D(I,J))  
0138      WRITE(6,87)  
0139      87 FORMAT(1X,"THE INTERSET DISTANCES")  
0140      DO 86 I=1,NC-1  
0141      DO 86 J=I+1,NC  
0142      WRITE(6,84)I,J,D(I,J)  
0143      84 FORMAT(1X,"I=",I1,2X,"J=",J1,2X,"D(I,J)=",E15.8)  
0144      86 CONTINUE  
0145      CALL CLOSE(CDCB)  
0146      999 STOP  
0147      END  
0148      END$
```

