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TECHNICAL REPORT STANDARD TITLE PAGE

1. Report No. 2261-10		2. Government Accession No.		3. Recipient's Catalog No.	
4. Title and Subtitle Kansas Environmental and Resource Study: A Great Plains Model-Use of Feature Extraction Techniques for the Texture and Context Information in ERTS Imagery		5. Report Date March 17, 1974		6. Performing Organization Code	
7. Author(s) R. M. Haralick and Robert Bosley		8. Performing Organization Report No.		10. Work Unit No.	
9. Performing Organization Name and Address University of Kansas Center for Research, Inc. c/o Space Technology Center 2291 Irving Hill Dr.-Campus West Lawrence, Kansas 66045		11. Contract or Grant No. NAS5-21822, Task 1		13. Type of Report and Period Covered Type III Report August 1, 1972-March 17, 1974	
12. Sponsoring Agency Name and Address Goddard Space Flight Center Greenbelt, Maryland 20771 GSFC Technical Monitor: G. R. Stonesifer, Code 430		14. Sponsoring Agency Code		15. Supplementary Notes	
16. Abstract A procedure is developed to extract cross-band textural features from ERTS multi-spectral scanner imagery. Evolving from a single-image texture extraction procedure which uses spatial dependence matrices to measure relative co-occurrence of nearest neighbor grey tones, the cross-band texture procedure uses the distribution of neighboring grey tone N-tuple differences to measure the spatial inter-relationships, or co-occurrences, of the grey tone N-tuples present in a texture pattern. In both procedures, texture is characterized in such a way as to be invariant under linear grey tone transformations. However, the cross-band procedure compliments the single-image procedure by extracting texture information and spectral information contained in ERTS multi-images. Classification experiments show that when used alone, without spectral processing, the cross-band texture procedure extracts more information than the single-image texture analysis. Results show an improvement in average correct classification from 86.2% to 88.3% for ERTS image no. 1021-16333 with the cross-band texture procedure. However, when used together with spectral features, the single-image texture plus spectral features perform better than the cross-band texture plus spectral features, with an average correct classification of 93.8% and 91.6% respectively.					
17. Key Words (Selected by Author(s)) Texture, land-use classification, spatial and spectral processing			18. Distribution Statement		
19. Security Classif. (of this report) UNCLASSIFIED		20. Security Classif. (of this page) UNCLASSIFIED		21. No. of Pages 249	22. Price*

KANSAS ENVIRONMENTAL AND RESOURCE STUDY:  
A GREAT PLAINS MODEL

SPECTRAL AND TEXTURAL PROCESSING OF ERTS IMAGERY

PREFACE

A procedure is developed to extract cross-band textural features from ERTS multi-spectral scanner imagery. Evolving from a single-image texture extraction procedure which uses spatial dependence matrices to measure relative co-occurrence of nearest neighbor grey tones, the cross-band texture procedure uses the distribution of neighboring grey tone N-tuple differences to measure the spatial interrelationships, or co-occurrences, of the grey tone N-tuples present in a texture pattern. In both procedures, texture is characterized in such a way as to be invariant under linear grey tone transformations. However, the cross-band procedure compliments the single-image procedure by extracting texture information and spectral information contained in ERTS multi-images. Classification experiments show that when used alone, without spectral processing, the cross-band texture procedure extracts more information than the single-image texture analysis. Results show an improvement in average correct classification from 86.2% to 88.8% for ERTS image no. 1021-16333 with the cross-band texture procedure. However, when used together with spectral features, the single-image texture plus spectral features perform better than the cross-band texture plus spectral features, with an average correct classification of 93.8% and 91.6%, respectively.

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KANSAS ENVIRONMENTAL AND  
RESOURCE STUDY: A GREAT PLAINS  
MODEL  
MARCH 1974

Type III Final Report for the Period  
August 1, 1972 - March 17, 1974

Prepared for:

National Aeronautics and Space Administration  
Goddard Spaceflight Center  
Greenbelt, Maryland 20771

Contract No. NAS 5-21822

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THE UNIVERSITY OF KANSAS CENTER FOR RESEARCH, INC.  
2385 Irving Hill Rd.—West Campus Lawrence, Kansas 66044

**KANSAS ENVIRONMENTAL AND RESOURCE STUDY:  
A GREAT PLAINS MODEL**

**Use of Feature Extraction Techniques for the Texture  
and Context Information in ERTS Imagery:**

**Spectral and Textural Processing of ERTS Imagery**

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**March 17, 1974  
Type III Final Report for the Period August 1, 1972 - March 17, 1974**

**Prepared for:**

**NATIONAL AERONAUTICS AND SPACE ADMINISTRATION  
GODDARD SPACEFLIGHT CENTER  
GREENBELT, MARYLAND 20771**

**Type III Final Report**  
**for the period: August 1, 1972 - March 17, 1974**

**Title of Investigation: Use of Feature Extraction Techniques for Texture Context  
Information in ERTS Imagery**

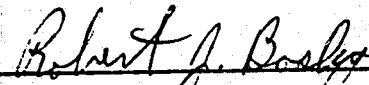
**ERTS-A Proposal Number: 60-1**

**Task Number: 1**

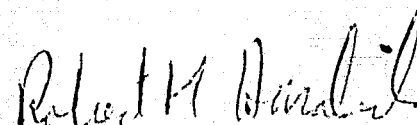
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## I. INTRODUCTION

A procedure is developed to extract textural features for automatic analysis of ERTS multi-spectral scanner imagery. Previous work (Haralick, 1973; Haralick, 1972) indicates that useful textural features can be computed from the co-occurrence matrices for grey tones in specific spatial relationships on an image. The performance of the land-use classification algorithm using these textural features from only one band is encouraging; up to 75 per cent of the images were correctly classified (Haralick, 1973). Since textural features and spectral features of ERTS multi-images provide different kinds of information, a significant increase in identification accuracy will occur when both features are used together.

Adoption of the texture procedure for multi-images leads to excessive amounts of storage for the grey tone N-tuple co-occurrence matrices. Therefore, to solve the storage problem we measure grey tone N-tuple differences instead of grey tone N-tuples and assume an ellipsoidally symmetric functional form for the co-occurrence distribution of multi-image grey tone N-tuple differences.

It is expected that the estimated parameters of the ellipsoidally symmetric distribution will lead to textural features that can distinguish between texturally distinct categories on ERTS MSS images over Kansas. In order to obtain more texture information, the dimensionality of the grey tone N-tuples was increased from the original four MSS bands by the addition of cross-band product terms for higher order components. This procedure for cross-band texture analysis of multi-images provides a natural extension of the single-image texture analysis while retaining its advantages: invariance under translating and scaling transformations, low storage requirements, and direct proportionality between the number of operations required to process an image and the number of resolution cells present in an image.



## II. TEXTURE

Texture and tone are two fundamental pattern elements used in the interpretation of image data. The concept of tone is concerned with the whiteness, greyness, or blackness of resolution cells of the image. The concept of texture is concerned with the spatial distribution of the grey tones. Tone is based upon the varying shades of grey of the resolution cells in the image, while texture is based upon the spatial distribution of grey tones. However, texture and tone are not independent concepts but are intrinsically related to one another. Although either property can dominate the other depending upon the image context, texture and tone are always present.

When one attempts to objectively use tone and texture pattern elements, the texture-tone concept must be explicitly defined. This can be visualized as follows. When a small area patch of an image has little variation of features of discrete grey tone, then that area is dominated by tonal properties. As the number of distinguishable features of discrete grey tone increases within the patch, then the texture properties will dominate. The size of the small area patch, the relative sizes of the discrete features and the number of distinguishable discrete features are all crucial in this distinction. When the size of the small area patch is reduced to one resolution cell, the only property present is tone. When there is no spatial pattern in the tonal features and the grey tone variation between features is wide, a fine texture results. And as the spatial pattern becomes more defined using more and more resolution cells, then a coarser texture results.

Texture can be termed as being fine, coarse, smooth, rippled, mottled irregular, or lineated. Texture is a property of nearly all surfaces, the grain of wood, the weave of fabric, the pattern of crops in a field, etc. Although texture is quite easy for humans to recognize and describe, it is quite subjective by its nature and is extremely difficult to precisely define and analyze by digital computers. Since the texture of images contains important information for discrimination purposes, textural features could be very useful.

### III. REVIEW OF PAST WORK ON TEXTURE

To date there has been at least six different approaches to the problems of measuring and characterizing texture of images: autocorrelation functions, optical transforms, digital transforms, edgeness, structural elements, and spatial grey tone co-occurrence probabilities. The first three approaches all measure spatial frequency either directly or indirectly. Spatial frequency is related to texture because fine textures are rich in high frequencies while coarse textures are rich in low frequencies.

One alternative approach to viewing texture as spatial frequency distribution is to view texture as the amount of edgeness per unit area. Fine textures have a high number of edges per unit area whereas coarse textures have a small number of edges per unit area.

The structural element approach uses a matching procedure to detect the spatial regularity of shapes called structural elements in a binary image. When the structural elements themselves are single resolution cells, the information provided by this approach is the autocorrelation function of the binary image. By using larger and more complex shapes, a more generalized autocorrelation can be computed.

The grey tone spatial dependence approach characterizes texture by the spatial distribution of its grey tones. In coarse textures the distribution changes only slightly with distance, but for fine textures it changes rapidly with distance.

Because of our familiarity with the concepts of spatial frequency and edgeness, these approaches to texture characterizations are readily employed. However, an inherent problem exists with these approaches in regard to grey tone calibration of the image and they are not invariant under even a linear grey tone translation. And the price paid for invariance by compensating with quantization is a loss of grey tone precision in the quantized image.

The power of the structural element approach is that it emphasizes the shape aspects of the discrete tonal features. Weakness of this approach lies in that it can only do so for binary images.

The power of the spatial grey tone co-occurrence approach lies in characterizing the spatial inter-relationships of the grey tones in a texture pattern in such a way that is invariant under monotonic grey tone transformations. Weakness of the approach lies in failure to capture the shape aspects of the discrete tonal features.

#### IV. TEXTURAL FEATURES

The above description of texture is an idealization of what actually occurs, a gross simplification. Discrete tonal features are actually quite subjective in that they do not necessarily stand out as entities by themselves. Therefore, the texture analysis presented here is concerned with more general or macroscopic concepts rather than discrete tonal features. The procedure developed by Haralick (Haralick, 1972) for obtaining the textural features of an image is based on the assumption that the texture information on an image  $I$  is contained in the overall spatial co-occurrence relationship which the greytone in the image  $I$  have to one another. More specifically, we assume that this texture information is adequately specified by a set of spatial grey tone dependence matrices, which are computed for various angular relationships and distances between neighboring resolution cell pairs on the image. All of the textural features are then derived from these angular nearest neighbor spatial grey tone dependence matrices.

##### IV.1 Spatial Grey Tone Dependence Matrices

Let  $G = \{0, 1, \dots, N_g\}$  be the set of possible grey tones that each resolution cell can take on after image normalization by equal probability quantizing to  $N_g$  levels. It can be shown that this quantization guarantees that images which are a monotonic transformation of one another, such as lighter or darker images due to variations in film, lighting, or development, will produce the same results. Let  $N_x$  be the number of resolution cells in the horizontal direction and  $N_y$  the number of resolution cells in the vertical direction in the image to be analyzed so that  $L_x = \{1, 2, \dots, N_x\}$  and  $L_y = \{1, 2, \dots, N_y\}$  are the horizontal and vertical spatial domains. Then  $L_y \times L_x$  will be the set of resolution cells of the image. And the image  $I$  can be represented as a function which assigns some grey tone in  $G$  to each resolution cell or pair of coordinates in  $L_y \times L_x$ ;  $I: L_y \times L_x \rightarrow G$ .

Essential to our conceptual framework of texture are four closely related measures called angular nearest neighbor grey tone spatial dependence matrices. The concept of angular nearest neighbor for a resolution cell is the adjacent resolution cell for a given angle, as shown in Figure 1.

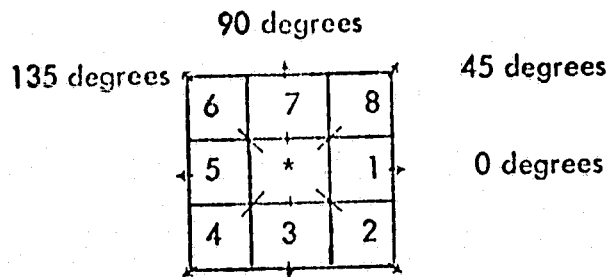


FIGURE 1. Eight nearest neighbor resolution cells of cell '\*'. Resolution cells 1 and 5 are the 0-degree nearest neighbors to resolution cell '\*', resolution cells 2 and 6 are the 135-degree nearest neighbors, etc. Note that this information is purely spatial, having nothing to do with grey tone values.

We assume that the texture information in our image  $I$  is contained in the overall or "average" spatial relationship which the grey tones in image  $I$  have to one another. Specifically, we shall assume that this information is adequately specified by the matrix of relative frequencies  $P_{ij}$  with which two neighboring resolution cells separated by distance  $d$  occur on the image, one with grey tone  $i$  and the other with grey tone  $j$ . These matrices of spatial grey tone dependence frequencies are a function of the angular relationship between the neighboring resolution cells as well as a function of the distance between them. Figure 2 illustrates the set of all horizontal neighboring resolution cells separated by distance 1. This set along with the image grey tones would be used to calculate a distance  $i$  horizontal spatial grey tone co-occurrence matrix. Formally, for angles quantized to  $45^\circ$  intervals the unnormalized frequencies are defined by:

$$P(i, j, d, 0^\circ) = \# \{ ((k, l), (m, n)) \in (L_y \times L_x) \times (L_y \times L_x) \mid k-m=0, |l-n|=d, I(k, l)=i, I(m, n)=j \}$$

$$P(i, j, d, 45^\circ) = \# \{ ((k, l), (m, n)) \in (L_y \times L_x) \times (L_y \times L_x) \mid (k-m=d, l-n=-d) \text{ or } (k-m=-d, l-n=d), I(k, l)=i, I(m, n)=j \}$$

$$P(i, j, d, 90^\circ) = \# \{ ((k, l), (m, n)) \in (L_y \times L_x) \times (L_y \times L_x) \mid |k-m|=d, l-n=0, I(k, l)=i, I(m, n)=j \}$$

$$P(i, j, d, 135^\circ) = \# \{ ((k, l), (m, n)) \in (L_y \times L_x) \times (L_y \times L_x) \mid (k-m=d, l-n=d) \text{ or } (k-m=-d, l-n=-d), I(k, l)=i, I(m, n)=j \}$$

Note that these matrices are symmetric;  $P(i, j; d, a) = P(j, i; d, a)$ . The distance metric  $\rho$  implicit in the above equations can be explicitly defined by

$$\rho((k, l), (m, n)) = \max \{ |k-m|, |l-n| \}.$$

(1,1)	(1,2)	(1,3)	(1,4)
(2,1)	(2,2)	(2,3)	(2,4)
(3,1)	(3,2)	(3,3)	(3,4)
(4,1)	(4,2)	(4,3)	(4,4)

$$L_y = \{1, 2, 3, 4\}$$

$$L_x = \{1, 2, 3, 4\}$$

$$\begin{aligned}
R_H &= \left\{ ((k,l), (m,n)) \in (L_y \times L_x) \times (L_y \times L_x) \mid k-m=0, \quad ||l-n|| = 1 \right\} \\
&= \left\{ ((1,1), (1,2)), ((1,2), (1,1)), ((1,2), (1,3)), ((1,3), (1,2)), \right. \\
&\quad ((1,3), (1,4)), ((1,4), (1,3)), ((2,1), (2,2)), ((2,2), (2,1)), \\
&\quad ((2,2), (2,3)), ((2,3), (2,2)), ((2,3), (2,4)), ((2,4), (2,3)), \\
&\quad ((3,1), (3,2)), ((3,2), (3,1)), ((3,2), (3,3)), ((3,3), (3,2)), \\
&\quad ((3,3), (3,4)), ((3,4), (3,3)), ((4,1), (4,2)), ((4,2), (4,1)), \\
&\quad \left. ((4,2), (4,3)), ((4,3), (4,2)), ((4,3), (4,4)), ((4,4), (4,3)) \right\}
\end{aligned}$$

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FIGURE 2. Illustrates the set of all Distance 1 Horizontal  
Neighboring Resolution Cells on a 4 x 4 Image.



For an example of the four distance 1 grey tone spatial dependence matrices, consider Figure 3. Figure 3-a represents a 4 x 4 image with four grey tones, ranging from 0 to 3. Figure 3-b shows the general form for any grey tone spatial dependence matrix. For example, the element in the (2,1)-st position of the distance 1 horizontal  $P_H$  matrix is the total number of times two grey tones of value 2 and 1 occurred horizontally adjacent to each other. To determine this number, we count the number of pairs of resolution cells in  $R_H$  such that the first resolution cell of the pair has grey tone 2 and the second resolution cell of the pair has grey tone 1. Figure 3-c through 3-f shows all four distance 1 grey tone spatial dependence matrices.

From the grey tone dependence matrices a set of 17 textural features is derived. The equations defining these 17 features are given in Appendix 1. To illustrate the significance of these features, three are defined as follows:

$$f_1 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} \left( \frac{P(i,j)}{\#R} \right)^2,$$

$$f_{12} = \sum_{n=0}^{N_g-1} n^2 \left\{ \sum_{|i-j|=n} \left( \frac{P(i,j)}{\#R} \right) \right\}$$

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$$f_3 = \frac{\sum_{i=1}^{N_g} \sum_{j=1}^{N_g} \frac{ij P(i,j)}{\#R} - \mu_x \mu_y}{\sigma_x \sigma_y}$$

where

$\#R$  = number of resolution cells pairs, and  $\mu_x$  and  $\sigma_x$  are the mean and standard deviation of the marginal distribution  $P_x$  defined by

$$P_x(i) = \sum_{j=1}^{N_g} \frac{P(i,j)}{\#R}$$

and  $\mu_y$  and  $\sigma_y$  are the mean and standard deviation of the marginal distribution  $P_y$  defined by:

$$P_y(j) = \sum_{i=1}^{N_g} \frac{P(i,j)}{\#R}$$

0	0	1	1
0	0	1	1
0	2	2	2
2	2	3	3

Figure 3-a.

		Grey Tone			
		0	1	2	3
Grey Tone	0	#(0,0)	#(0,1)	#(0,2)	#(0,3)
	1	#(1,0)	#(1,1)	#(1,2)	#(1,3)
	2	#(2,0)	#(2,1)	#(2,2)	#(2,3)
	3	#(3,0)	#(3,1)	#(3,2)	#(3,3)

Figure 3-b. This shows the general form of any grey tone spatial dependence matrix for an image with integer grey tone values 0 to 3. #(i,j) stands for number of times grey tones i and j have been neighbors.

$$0^\circ \quad P_H = \begin{pmatrix} 4 & 2 & 1 & 0 \\ 2 & 4 & 0 & 0 \\ 1 & 0 & 6 & 1 \\ 0 & 0 & 1 & 2 \end{pmatrix}$$

Figure 3-c.

$$90^\circ \quad P_V = \begin{pmatrix} 6 & 0 & 2 & 0 \\ 0 & 4 & 2 & 0 \\ 2 & 2 & 2 & 2 \\ 0 & 0 & 2 & 0 \end{pmatrix}$$

Figure 3-d.

$$135^\circ \quad P_{LD} = \begin{pmatrix} 2 & 1 & 3 & 0 \\ 1 & 2 & 1 & 0 \\ 3 & 1 & 0 & 2 \\ 0 & 0 & 2 & 0 \end{pmatrix}$$

Figure 3-e.

$$45^\circ \quad P_{RD} = \begin{pmatrix} 4 & 1 & 0 & 0 \\ 1 & 2 & 2 & 0 \\ 0 & 2 & 4 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

Figure 3-f.

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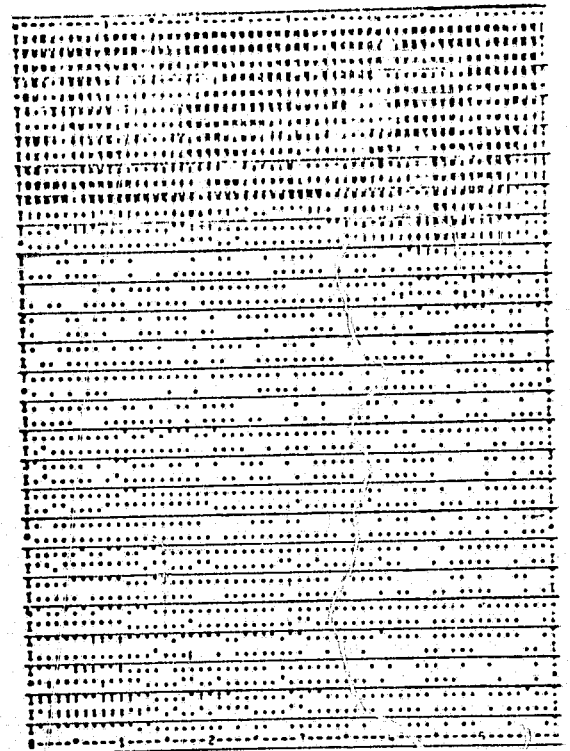
To explain the significance of these features, let us consider the kind of values they take on two different land use category images. Figure 4 shows the digital printout of two sub-images of size  $64 \times 64$  resolution cells (approximately 8.5 square mile area) from MSS band 5 of 1002-18134, see Figure 10. The image shown in 4 (a) belongs to the grass land category and the image in Figure 4(b) is mostly water. Values of the features  $f_1$ ,  $f_{12}$ , and  $f_3$  are also shown for these images in Figure 4.

The angular second moment feature (ASM),  $f_1$ , is a measure of homogeneity of the image. In a homogeneous image, such as shown in 4(b), there are very few dominant grey tone transitions. Hence, the P matrix for this image will have fewer entries of large magnitude. For an image like the one shown in Figure 4(a), the P matrix will have a large number of small entries and hence the ASM feature which is the sum of squares of the entries in the P matrix will be smaller. A comparison of the ASM values given below the images in Figure 4 shows the usefulness of the ASM feature as a measure of the homogeneity of the image.

The contrast feature,  $f_{12}$ , is obtained as a difference moment of the P matrix and is a measure of the contrast or the amount of boundaries present in an image. Since there is a large amount of boundaries present in the image 4(a) compared to the image shown in 4(b), the contrast feature for the grassland image has consistently higher values compared to the water body image.

The correlation feature,  $f_3$ , is a measure of linear grey tone dependencies in the image. For both the images shown in Figure 4, the correlation feature is somewhat higher in the horizontal ( $0^\circ$ ) direction, along the line of scan. The water body image consists mostly of a constant grey tone value for the water plus some additive noise. Since the noise samples are mostly uncorrelated, the correlation features for the water body image have lower correlation values compared to the grassland image. Also the grassland image has a considerable amount of linear structure along  $45^\circ$  lines across the image and hence the value of the correlation feature is higher along this direction compared to the values for  $90^\circ$  and  $135^\circ$  directions.

The various features presented here are all functions of distance and angle. The angular dependencies present a special problem. Suppose image A has features  $a, b, c, d$  for angles  $0^\circ, 45^\circ, 90^\circ,$  and  $135^\circ$  and image B is identical to A except that B is rotated  $90^\circ$  with respect to A. Then B will have features  $c, d, a, b,$  for angles  $0^\circ, 45^\circ, 90^\circ,$  and  $135^\circ$  respectively. Since the texture



a. Grassland

b. Water Body

Angle	ASM	Contrast	Correlation	ASM	Contrast	Correlation
0°	.0128	3.048	.8075	.1016	2.153	.7254
45°	.0080	4.011	.6366	.0771	3.057	.4768
90°	.0077	4.014	.5987	.0762	3.113	.4646
135°	.0064	4.709	.4610	.0741	3.129	.4650
Avg.	.0087	3.945	.6259	.0822	2.863	.5327

Figure 4. Textural Features for Two Different Land Use Category Images.

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context of A is the same as the texture context of B, any decision rule using the angular features a, b, c, d must produce the same results for c, d, a, b, or for that matter b, c, d, a (45° rotation) and d, a, b, c, (135° rotation). To guarantee this, we do not use the angularly dependent features directly. Instead, we merge the four arrays by summing corresponding elements. The merged array is then used for computing the 17 texture features defined in Appendix I.

#### IV.2 Textural Features for Multi-Images

Results of previous work in texture using the spatial grey tone dependence matrices as the basis from which all textural features are extracted has been very encouraging (Haralick, 1973). The good performance of these texture features can be seen by the results of the classification experiments. The computational aspects of the procedure are also notable. The number of operations required to process an image using the spatial grey tone dependence matrices is directly proportional to the number of resolution cells, N, present in an image. In comparison, the number of operations needed to use Fourier or Hadamard transforms to extract texture information are of the order of  $N \log N$ . And, to compute the entries in the spatial grey tone dependence matrices, one needs to keep only two lines of image data in core at a time, keeping storage requirements to a minimum.

Even with these advantages, however, the extraction of texture information from multiimages, as in the case of ERTS MSS data, forces a new approach to the measurement of grey tone N-tuple co-occurrences. The use of the spatial dependence matrices requires that they be stored in the computer. For multi-images containing grey tone N-tuples, we have too many possible grey tone N-tuples which can neighbor each other and as a result, the dependence matrices will be very large. For example, for four MSS bands in which each grey tone can range through 64 levels, each matrix would have  $64^4 \times 64^4$  elements. Even using the symmetry of the matrices to reduce the number of entries does not help since there would be on the order of  $10^{15}$  entries.

The spatial dependence matrices, however, provide a way of escape. In using these matrices, it was observed that they are heavily weighted along the diagonal with decreasing entries farther from the diagonal. Figure 5 gives an



example of one of these matrices. Note that the number of entries decreases as we move away from the diagonal. This suggests that neighboring resolution cells are similar. Choosing any resolution cell in an image at random, we are very likely to find nearly identical neighbors to the cell in all directions and less likely to find dissimilar neighbors. Clearly, a measure which indicates how similar the neighboring N-tuples are and how fast the similarity drops off with distance must contain textural information about the object imaged.

It is therefore reasonable to measure the difference between neighboring grey tone N-tuples and observe this distribution instead of computing the number of times each N-tuple neighbors every other N-tuple. In both cases we measure the co-occurrence of nearest neighbor grey tone N-tuples.

Since the textural features are based on the spatial dependence of grey tone N-tuples, our first step must be to define a binary relation between neighboring resolution cells on which the co-occurrence of grey tone N-tuples can be counted. As above, let  $L_x = \{1, 2, \dots, N_x\}$  and  $L_y = \{1, 2, \dots, N_y\}$  be the set of column and row indexes, respectively, so that  $L_y \times L_x$  is the set of resolution cells in the image. Let  $G = \{0, 1, \dots, N_g\}$  be the set of possible grey tones that each component of every grey tone N-tuple can be assigned. Then, the image I can be defined by  $I: L_y \times L_x \rightarrow G \times G \times \dots \times G$ .

Let R be the set of all pairs of resolution cells in a specified spatial relation. Then R is a binary relation on the set  $L_y \times L_x$ ;  $R \subseteq (L_y \times L_x) \times (L_y \times L_x)$ . For example, the set of all distance 1 horizontally neighboring pairs of neighboring resolution cells would be defined by:

$$R = \left\{ \left( (k, l), (m, n) \right) \in (L_y \times L_x) \times (L_y \times L_x) \mid k-m=0, |l-n|=1 \right\}.$$

The co-occurrence frequency of grey tone N-tuples  $(i_1, i_2, \dots, i_N)$  and  $(j_1, j_2, \dots, j_N)$  in spatial relation defined by R is

$$P \left( (i_1, \dots, i_N), (j_1, \dots, j_N) \right) = \frac{\# \left\{ \left( (k, l), (m, n) \right) \in R \mid I(k, l) = (i_1, \dots, i_N), I(m, n) = (j_1, \dots, j_N) \right\}}{\#R}$$

where # denotes the number of elements in the set.

Note that this R is symmetric. Assume that  $\left( (k, l), (m, n) \right)$  is in R. Then  $k-m=0$ , and  $|l-n|=1$  from the definitions of R. But  $|l-n|=1$  when  $|n-l|=1$ .



And if  $|n-1|=1$  and  $k-m=0$ , then  $((m, n), (k, l))$  is in  $R$ . Thus,  $R$  is symmetric. In fact, by the symmetry of any distance function,  $R$ , in general, must be symmetric. And since  $R$  is symmetric,  $P$  is also symmetric.

### IV.3 Textural Feature Extraction Procedure

Let  $R$  be a symmetric binary relation pairing nearby neighboring resolution cells. We define the frequency of grey tone  $N$ -tuple differences co-occurring in the spatial configuration defined by  $R$  as

$$P(x_1, x_2, \dots, x_N) = \frac{\# \left\{ ((i, j), (m, n)) \in R \mid I(i, j) - I(m, n) = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} \right\}}{\#R}$$

Note that  $P$  is an even function since

$$\begin{aligned} P(x_1, x_2, \dots, x_N) &= \# \left\{ ((i, j), (m, n)) \in R \mid I(i, j) - I(m, n) = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} \right\} / \#R \\ &= \# \left\{ ((i, j), (m, n)) \in R^{-1} \mid I(i, j) - I(m, n) = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} \right\} / \#R \\ &= \# \left\{ ((m, n), (i, j)) \in R \mid I(i, j) - I(m, n) = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} \right\} / \#R \\ &= \# \left\{ ((m, n), (i, j)) \in R \mid I(m, n) - I(i, j) = \begin{pmatrix} -x_1 \\ \vdots \\ -x_N \end{pmatrix} \right\} / \#R \\ &= P(-x_1, -x_2, \dots, -x_N). \end{aligned}$$

Referring to the monotonic behavior of nearly every column in the matrices of Figure 5, and assuming that this behavior occurs on every band of the ERTS multi-images, it is reasonable to assume that the even frequency distribution  $P(x_1, \dots, x_N)$  of the nearby grey tone  $N$ -tuple differences can be adequately approximated using an ellipsoidally symmetric distribution; thus we may write

$$P(x_1, x_2, \dots, x_N) = f(x'Ax)$$

for some monotonically decreasing function  $f$ .

This implies that only the function  $f$  and the matrix  $A$  need to be determined. We take  $f$  to be one of the two forms  $e^{-1/2 \mu^2}$ ,  $(1+\mu^2)^{-m}$ . Figure 6 is a scattergram of the differences for the first two bands of distance 1 horizontally neighboring resolution cells of a 64x64 sample image. Figures 6 and 7 clearly show the ellipsoidally symmetric functional form of the distribution of neighboring differences. In Appendix II  $N$ -dimensional spherical coordinate systems and ellipsoidally symmetric distributions are discussed and it is shown that the matrix  $A$  is proportional to the inverse covariance matrix of the  $N$ -tuple differences. Thus, we estimate  $A$  by a matrix proportional to the inverse of the estimate for the covariance matrix.

Therefore, if the image is blocked into subimages of small area so that each subimage is essentially of one category, we can expect the distribution of grey tone  $N$ -tuple differences over each subimage to be a function only of the assumed form of the function  $f$  and the covariance matrix of the difference vectors for grey tone  $N$ -tuples in a specified spatial relationship within the subimage. This leads us to consider textural features for multi-images based upon the elements of this spatial-spectral covariance matrix.

Consider each covariance matrix as a vector. Consider the distribution of the set of covariance matrices from the blocked image. Since the entries of the covariance matrix are the parameters of the distribution, we would like to have these entries invariant with respect to scale changes on the grey tone  $N$ -tuple differences. In order to do this, we scale the grey tone  $N$ -tuple differences so that all components have variance 1. The covariance matrix of these normalized differences is equivalent to the correlation matrix. Appendix III shows that this normalization procedure makes the covariance matrix invariant with respect to translating and scaling transformations on the grey tone  $N$ -tuples. The normalized covariance matrix can be considered as an extracted texture feature vector in an  $N(N-1)/2$  dimensional hyperspace.

Initial classification experiments indicated a need for more textural information and in order to provide this the dimensionality,  $N$ , of each resolution cell was increased from the four provided by the four MSS bands to eight by appending higher order terms and cross-band product terms for each cell. The resultant increase in correct classification accuracy can be seen in section VII.

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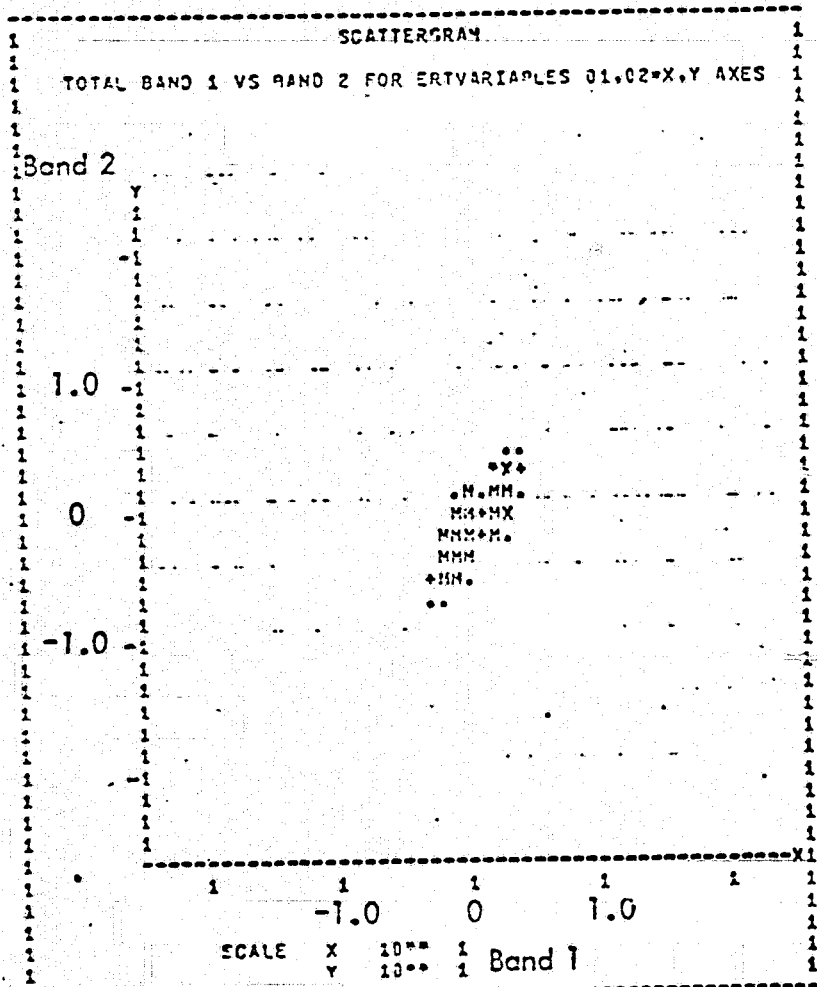


FIGURE 6. Scattergram showing ellipsoidally symmetric distribution of differences for Bands 4 and 5 over a sample 64 x 64 ERTS image.

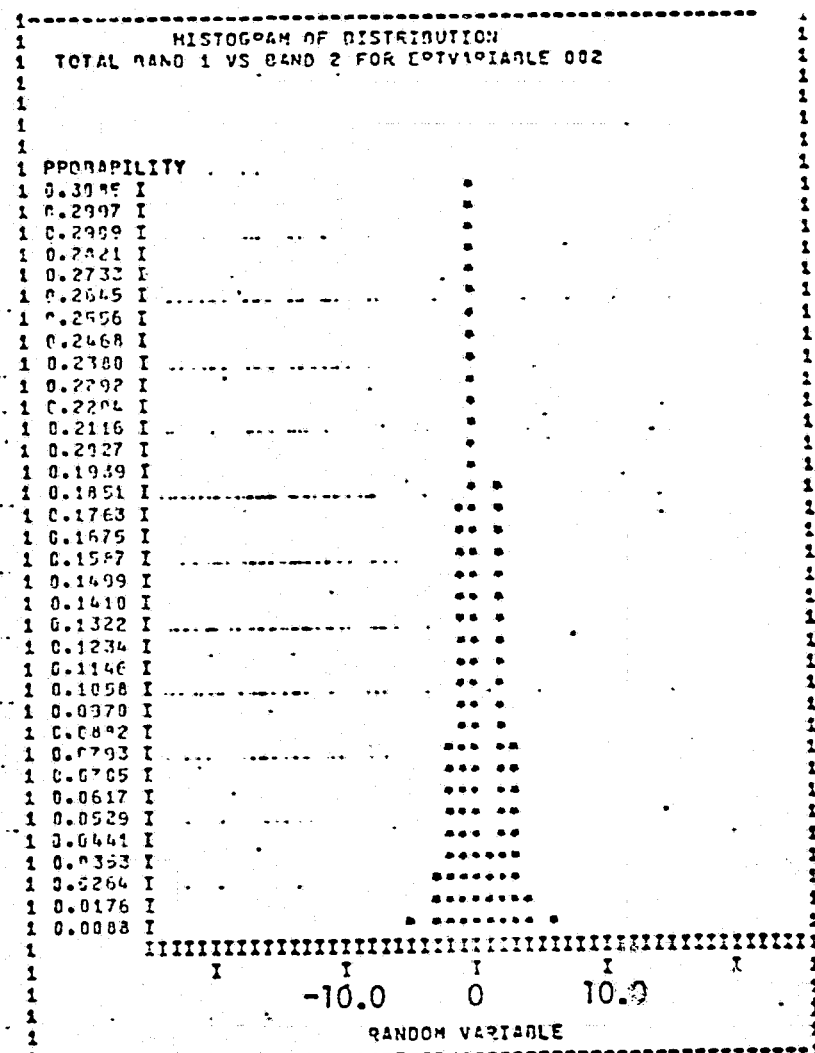


FIGURE 7. Histogram of the distribution of differences on Band 5 for distance 1 horizontally neighboring resolution cells.

## V. GROUND TRUTH ASSIGNMENT PROCEDURE

For all classification experiments, the only method available to determine ground truth was based upon the 9 inch by 9 inch ERTS image transparency. Initially, the land-use category names were determined with the aid of two photo interpreters and, when available, contour maps published by the U. S. Geological Survey.

After the categories to be used were decided upon, each ERTS image was processed. The image was blocked into 64 x 64 subimages for texture analysis on MSS band 5 and 32 by 32 subimages for cross-band texture analysis of MSS bands 4 thru 7. The ERTS Retrieval Program was then used to printout a picture of the same area that was processed. This was subdivided into 64 x 64 subimages. In this way, the location of each subimage was known and could be fairly accurately determined on a 7 inch by 7 inch print. Then, on the basis of its location, the subimage is assigned a ground truth category. Figure 8 shows a picture printout by the ERTS Retrieval Program for a portion of image 1021-16333 (see Figure 12) over Kansas City. The picture has been blocked into 64 x 64 subimages and by referring to the original image in Figure 12, each subimage was assigned a ground truth category.

Figures 8 and 12 also illustrate the major problem with ground truth assignment. This is the determination of ground truth when a subimage covers more than one ground truth category. The irregular boundary around the urban area of Kansas City makes ground truth assignment difficult. The deciding factor was the amount of area within the subimage from each category. The subimage was assigned to the category which had the largest area within its boundary. For this reason the accuracy with which the location of every subimage is determined becomes very important, and the picture printouts from the retrieval programs aided in this.

Since the starting and ending image row and column coordinates were known, the Retrieval Program could use the same coordinates for the picture printout. This printout could then be accurately divided into 64 by 64 subimages. The actual location of these subimages could then be determined fairly well, although it is sometimes difficult to locate objects in the printout that are on the original image. To aid in this, a few of the well defined objects were used to construct a 64 by 64 grid on the 7 inch by 7 inch

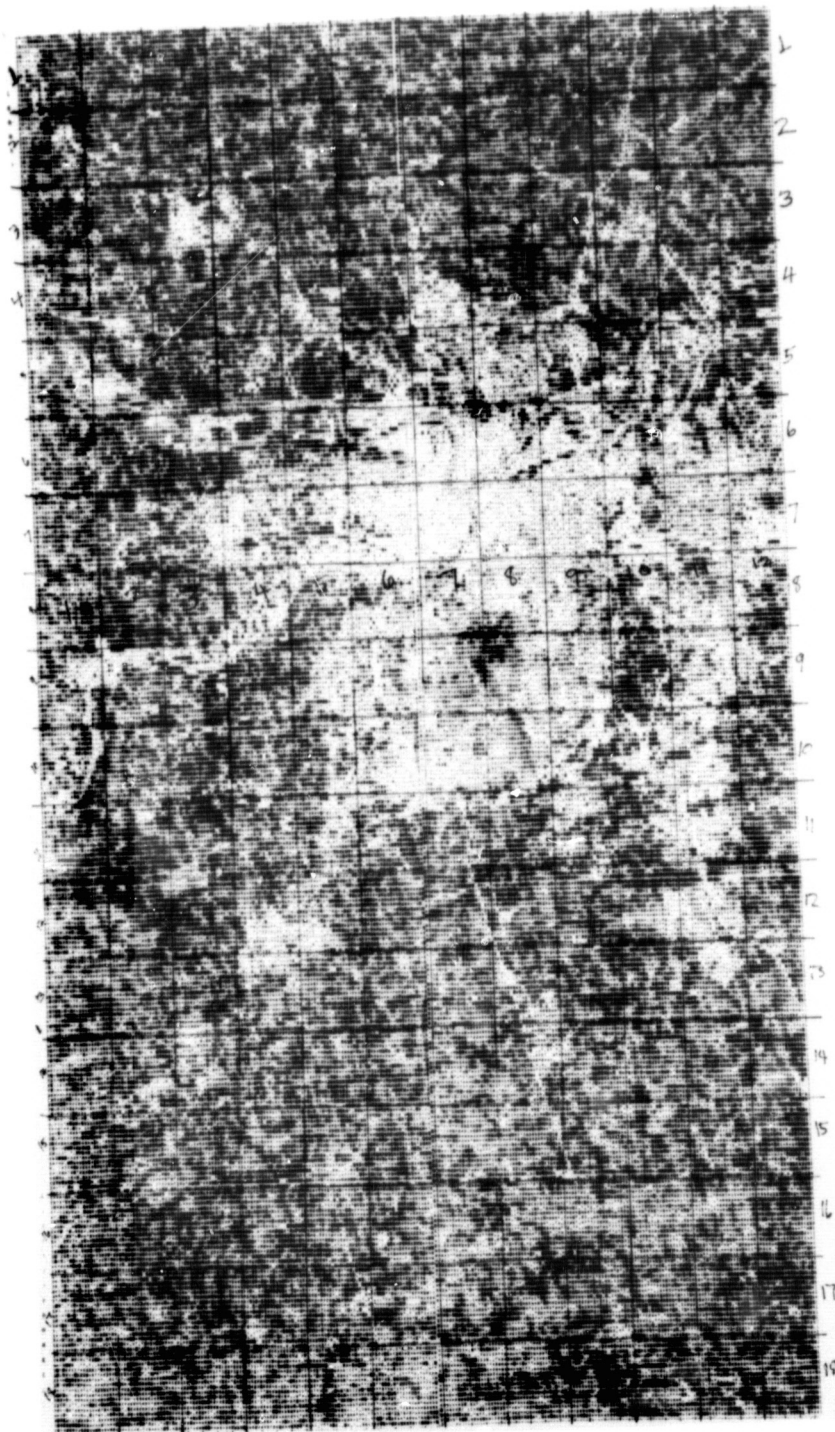


Figure 8. Printout of a Portion of Image 1021-16333 Over Kansas City by the ERTS Retrieval Program for Use in Ground Truth Assignment.

print. When there was a question in the ground truth, the printout could be referred to for the more accurate subimage location. The size of the subimage also relates directly to the accuracy with which its location on the print can be determined. It was found that 32 lines by 32 points per line was about as small as a subimage can be for good accuracy in ground truth assignment.

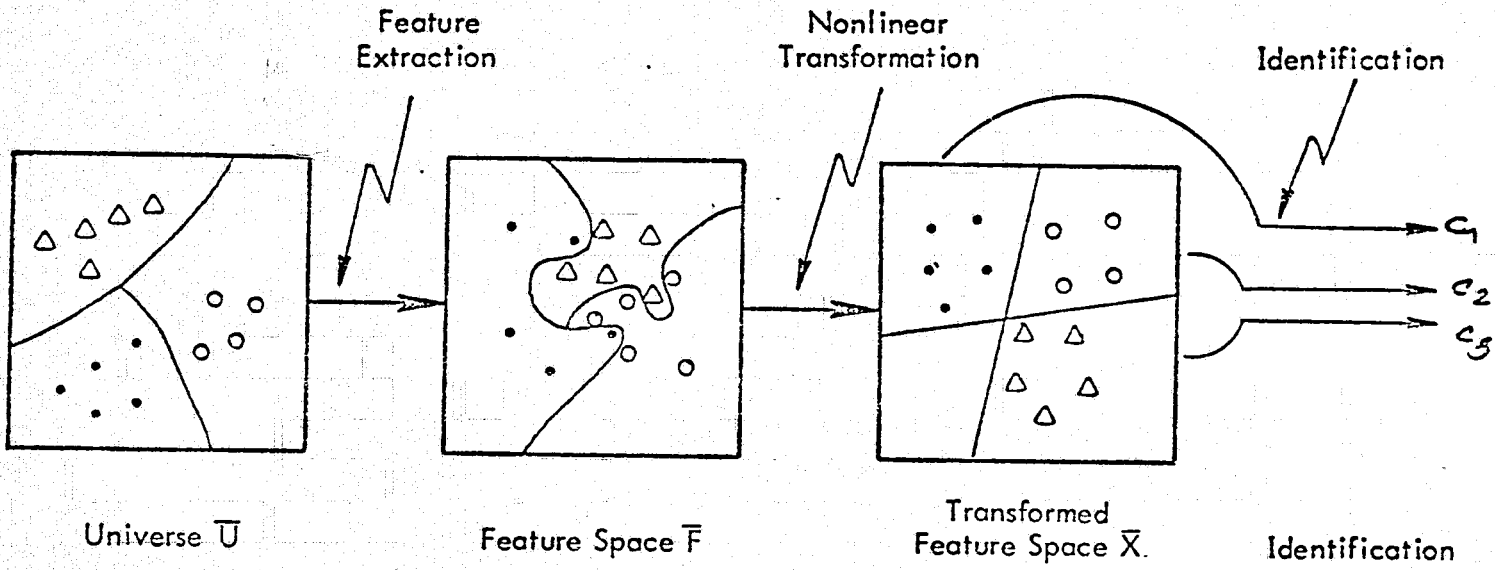
## VI. IDENTIFICATION PROCEDURE

In the classification experiments two different classification algorithms were used, a Bayes classifier and a piecewise linear classifier. The Bayes classifier assumes a multivariate normal distribution and randomly chooses pattern vectors for the training set where the piecewise linear classifier does not.

The problem of developing procedures for categorizing environmental units consists of the following.

With reference to Figure 9, the Universe  $\bar{U}$  consists of environmental units (for example rocks)  $U_1, U_2, \dots, U_T$  which belongs to one of  $R$  possible categories  $C_1, C_2, \dots, C_R$  (different land use categories). Of the large number of environmental units present in the universe, we observe a smaller subset of units  $U_1, U_2, \dots, U_N$ . Our observations consist of a set of measured values of  $n$  features  $f_1, f_2, \dots, f_n$  for each unit  $U$  sampled. Based on the information contained in the feature vectors  $F_1, F_2, \dots, F_N$ , the categories of the environmental units which produce these measurements being known, we want to develop an algorithm to identify the categories of new units based on the measurements they produce.





$$\bar{U} = \{U_1, U_2, \dots, U_T\}$$

$$\bar{F} = \{F_1, F_2, \dots, F_T\}; \quad F = [f_1 \ f_2 \ \dots \ f_n]^T$$

$$\bar{X} = \{X_1, X_2, \dots, X_T\}; \quad X = [x_1 \ x_2 \ \dots \ x_m]^T$$

The vectors  $F_i$  and  $X_i$  are usually referred to as feature vector and pattern vector respectively.

Figure 9. Identification Scheme.

The decision rule which assigns categories based on the values of features may be implemented in the feature space  $\bar{F}$  by partitioning  $\bar{F}$  into various regions and assigning categories to new units based on the regions to which their feature vectors belong. Efficient partitioning of the feature space may require complicated nonlinear decision boundaries (discriminant functions). Instead of deriving a decision rule in the feature space  $\bar{F}$ , we may transform the feature vectors into a new space  $\bar{X}$  and implement a decision rule in the new space  $\bar{X}$ . By using appropriate nonlinear transformations, we may be able to implement nonlinear decision boundaries in  $\bar{F}$  as linear decision boundaries in  $\bar{X}$ . Several procedures are available for deriving linear decision boundaries for partitioning  $\bar{X}$  into various regions, based on the information contained in a set of sample patterns  $X_1, X_2, \dots, X_N$  whose categories are known.

Identification Algorithms: In a widely used algorithm (Fukunaga 1972, Fu and Mendel 1970, Miesel 1972), the pattern space  $\bar{X}$  is separated into a number of regions using a set of hyperplanes (decision boundaries) whose locations are determined by the sample patterns. Each region is dominated by sample patterns of a particular category. When a new pattern is presented for identification, it is assigned a category depending on the region in which it belongs. If the new pattern  $X$  is located in a region dominated by sample patterns of category  $c_j$ , then  $X$  is classified as coming from category  $c_j$ .

To illustrate the procedure for obtaining the hyperplanes, consider the problem of separating the sample patterns  $X_1, X_2, \dots, X_{n_i}$  belonging to category  $c_i$  and  $X_{n_i+1}, X_{n_i+2}, \dots, X_{n_i+n_j}$  belonging to category  $c_j$ . We can write the linear discriminant function (hyperplane) which separates the patterns belonging to categories  $c_i$  and  $c_j$  as

$$h_{ij}(X) = V_{ij}^T X + v_{ij}^0 \geq 0 \text{ for } X \in c_i,$$

$$h_{ij}(X) = V_{ij}^T X + v_{ij}^0 < 0 \text{ for } X \in c_j$$

The vector  $V_{ij}$  and the scalar  $v_{ij}^0$  are to be determined from the information contained on the sample patterns.

If we introduce a new form to express the pattern vectors as

$$Z = [+1 \ x_1 \ x_2 \ \dots \ x_n]^T \text{ for } X \in c_i,$$

$$Z = [-1 \ -x_1 \ -x_2 \ \dots \ -x_n]^T \text{ for } X \in c_j$$

then the discriminant function can be written as

$$h_{ij}(Z) = W_{ij}^T Z = 0 \quad (1)$$

where  $W_{ij}$  is referred to as a weight vector and

$$h_{ij}(Z) = W_{ij}^T Z = 0$$

is the equation of a hyperplane in the transformed feature space.

The weight vector  $W_{ij}$  is chosen so as to satisfy equation 1 for as many training patterns as possible. Usually we do not know the precise form of  $h_{ij}$ . But, given our knowledge of the categories of the training patterns, we can postulate reasonable values  $g_{ij}(Z_k)$  for  $h_{ij}(Z_k)$  and choose  $W_{ij}$  to minimize the mean square error given by

$$\epsilon^2 = \frac{1}{n_i + n_j} \sum_{k=1}^{n_i + n_j} (W_{ij}^T Z_k - g_{ij}(Z_k))^2$$

Usually  $g_{ij}(Z_k)$  is taken to be +1 for  $k = 1, 2, \dots, n_i + n_j$ . We can rewrite  $\epsilon^2$  as,

$$\epsilon^2 = \frac{1}{(n_i + n_j)} [W_{ij}^T Y - G_{ij}^T] [Y^T W_{ij} - 1] \quad (2)$$

where  $Y = [Z_1 \ Z_2 \ \dots \ Z_{n_i + n_j}]$ , and

$$G_{ij} = [g_{ij}(Z_1) \ g_{ij}(Z_2) \ \dots \ g_{ij}(Z_{n_i + n_j})]$$

The weight vector which minimizes  $\epsilon^2$  given in equation 2 is given by

$$W_{ij} = (YY^T)^{-1} Y G_{ij}$$

which is the well-known normal equation set from linear least square theory.

For the multicategory problem involving  $N_R$  categories, a total of  $N_R(N_R - 1)/2$  hyperplanes must be determined using the procedure described above. After the hyperplanes are determined, the classification of new patterns is done as follows. For each category  $c_i$ , the number of hyperplanes,  $V_i$ , which give a positive response when the new pattern  $X$  is presented are determined using

$$V_i = \sum_{\substack{j=1 \\ j \neq i}}^{N_R} \frac{|W_{ij}^T Z| + W_{ij}^T Z}{2|W_{ij}^T Z|} ; i = 1, 2, \dots, N_R$$

where  $Z = \begin{bmatrix} 1 \\ X \end{bmatrix}$ .

X is assigned to category  $c_j$  if

$$V_j = \max_i \{V_i\}$$

If there is a tie between categories  $c_m$  and  $c_n$ , then X is assigned to  $c_m$  if  $W_{mn}^T Z \geq 0$  or to  $c_n$  if  $W_{mn}^T Z < 0$ . Several modifications of the linear discriminant function method and a multitude of other classification procedures may be found in the references cited.

## VII. RESULTS OF CLASSIFICATION EXPERIMENTS

Tables 1 thru 24 show the results of the classification experiments. Unless otherwise stated, the contingency tables are determined using the piecewise linear classification programs (RCLASS) given in Appendix IV. In each contingency table the number of errors (#ERR) and percent error (%ERR) is shown for both errors of commission and errors of omission. The final entry in the percent error column is the average of the percent error for each category. The average correct classification is simply the number of correct classifications divided by the number of incorrect classifications. The final column in each contingency table (%SD) is an estimate for the standard deviation of the probability for correct classification. This was taken from Afarani (Afarani, 1972) where he gives an estimate for the variance of the probability of correct classification when the sample size is fixed as:

$$\hat{\sigma}^2 = \frac{\left( \sum_i n_{ii} \right) \left( \sum_{\substack{i=j \\ i \neq j}} n_{ij} \right)}{n^3} = \frac{\left( \begin{matrix} \text{no.} \\ \text{correct} \end{matrix} \right) \left( \begin{matrix} \text{no.} \\ \text{errors} \end{matrix} \right)}{(\text{total no.})^3}$$

where  $n$  is the fixed sample size and  $n_{ij}$  is the number of classifications of units assigned to category  $j$  whose true category is  $i$ . Used directly, this gives an estimate over all categories for the entire contingency table. This is the last entry in the percent standard deviation (%SD) column. The remaining entries were determined by fixing  $i$  and estimating the standard deviation for each category  $i$ . This estimate says that as the number of samples tested increases and the number of correct classifications increase, the variance of the probability of correct classification decreases, as one would expect.

In order to obtain an initial estimate of performance of the multi-image texture features an experiment was performed on ERTS satellite imagery over Monterey Bay, California, image number 1002-18134 (see Figure 10) taken on July 25, 1972. Using a small set of 64 sampled  $32 \times 32$  subimages and training on 34 of these, 80 per cent of the remaining 30 test samples were correctly classified according to four land-use categories: coastal forest, annual grassland, urban area, and water, as shown in Table 1. This is encouraging since previous accuracy using spatial dependence matrices on band 5 with  $64 \times 64$  subimages over the same general area was only 70.5 per cent as shown in Table 2 (Haralick, 1973).

The ability to obtain good ground truth and several distinct categories in the California data was not the case for an ERTS image over Finney County, Kansas, which was used in later experiments. Approximately a 40 mile by 60 mile section near Garden City, Kansas, on image number 1330-16515 (see Figure 11), taken on June 18, 1973, was processed with initially four categories: grassland, large fields, small fields, and water. Both texture procedures, using the multi-image texture features with  $32 \times 32$  subimages and the single-image texture analysis on MSS band 5 with  $64 \times 64$  subimages, were used on the image. Tables 3 and 4 show the results of classification for distance 1 resolution cells while Tables 5 and 6 show distance 8 results. In both cases the single image classification is higher. However, when both distances 1 and 8 are used together, classification accuracy for both procedures is nearly identical, as shown in Tables 7 and 8, about 70 per cent. Tables 9 and 10 show results using the Bayes Classifier.

This implies that more information is contained in the single-band texture features than the multi-image texture features. In order to add more texture information, a measure of entropy (Kullback, 1959), given by

$$E = -\frac{1}{2} \log |P|$$

where  $P$  is the correlation matrix, was added to the cross-band texture feature set. Also, higher order components were appended to each grey tone  $N$ -tuple by squaring the grey tones and getting cross-band product terms. Only a few of these were added, increasing the grey tone  $N$ -tuple dimensionality from 4 to 8, which results in an increase in the number of feature vector components (elements in the correlation matrix) from 6 to 28 plus the entropy measure. The eight components in each grey tone  $N$ -tuple are: MSS Band 5, Band 6, Band 7,  $(\text{Band } 5)^2$ ,  $(\text{Band } 6)^2$ ,  $(\text{Band } 7)^2$ ,  $(\text{Band } 5) \times (\text{Band } 6)$ ,

(Band 5) x (Band 7). Figure 13 gives an illustration of a correlation matrix with the feature vector component designation to be used in indicating the components selected for input to the classification programs.

Table 11 shows the contingency table for the cross-band texture features using 9 of the 29 components (1,2,5,6,9,12, 20, 24, and 27 of figure 13) for a portion of the Garden City, Kansas, data. The increase in identification accuracy between the large and small fields results in an increase in overall correct classification, up to 87.1%.

The final classification experiment to test the cross-band texture analysis was made on ERTS image 1021-16333 (see Figure 12) taken on August 13, 1973, over Kansas City. Four land-use categories were chosen: cropland (directly north and south of Kansas City), urban area (Kansas City, Topeka), grassland (southwestern corner of the image), and water (Perry reservoir plus several small lakes). These areas were processed three separate ways:

- 1) spectrally, 32 by 32 subimages
- 2) texturally, 64 by 64 subimages on band 5
- and 3) cross-band texturally, 32 by 32 subimages.

Because of problems with ground truth and a small data set for water, that category was later dropped. The spectral processing involved obtaining the average grey tone over the subimage for each spectral band, giving 4 components for each feature vector. The textural processing was over a larger subimage than either the spectral or the cross-band textural processing. This was chosen because the 64 by 64 subimages have performed better in the past than the 32 by 32 subimages for the single-image texture analysis. The smaller subimage size was chosen for the cross-band texture processing so that the subimage would more likely be from only one category. The cross-band texture method uses spectral information which the single-image texture procedure does not have available. The smaller subimage size was also chosen for the spectral processing in order to provide an estimate of the amount of spectral information contained in the four MSS bands that is available to the cross-band texture analysis.

Single-image texture analysis was done for distance 1 nearest neighbors at all angles and all of the 17 texture features defined in Appendix I were computed for each subimage. The cross-band texture processing was also done for distance 1 using horizontally adjacent nearest neighbor grey tone N-tuples.

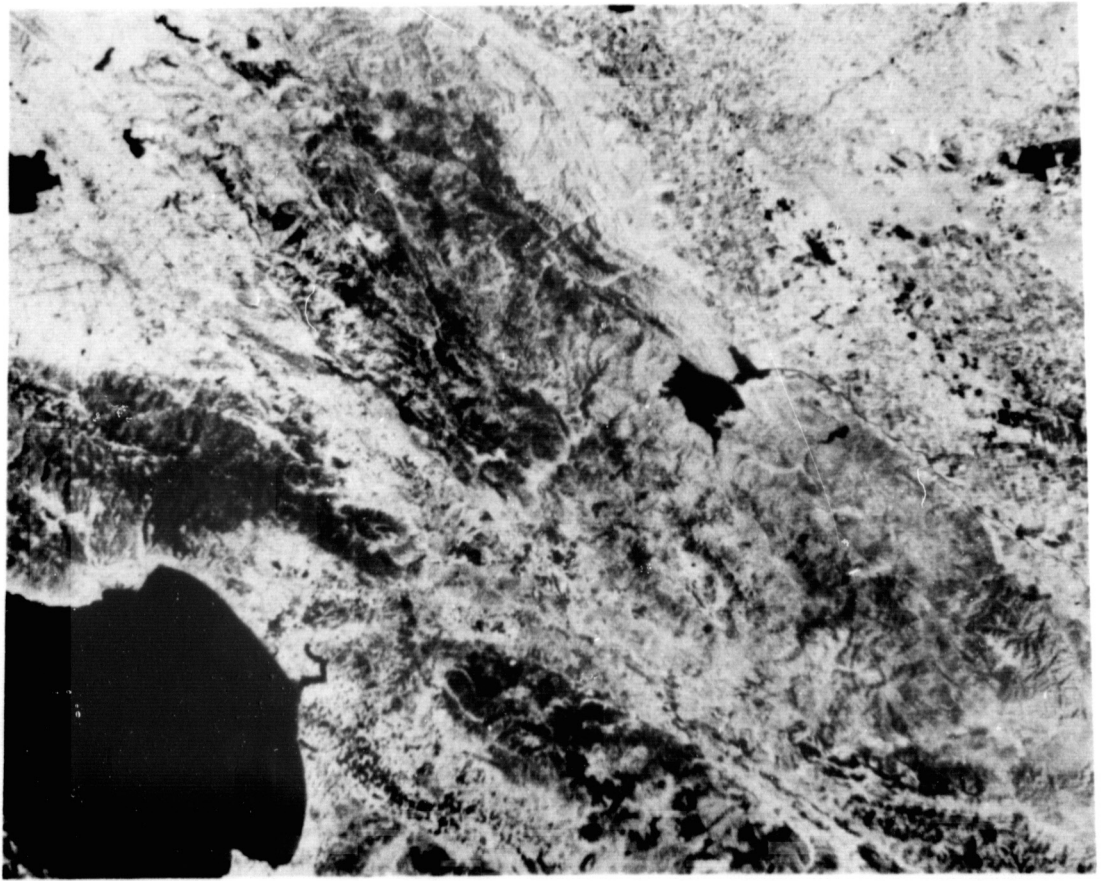


Figure 10. Part of ERTS Image No. 1002-18134 (MSS Band 5) Over Monterey Bay, California, Taken on July 25, 1972.

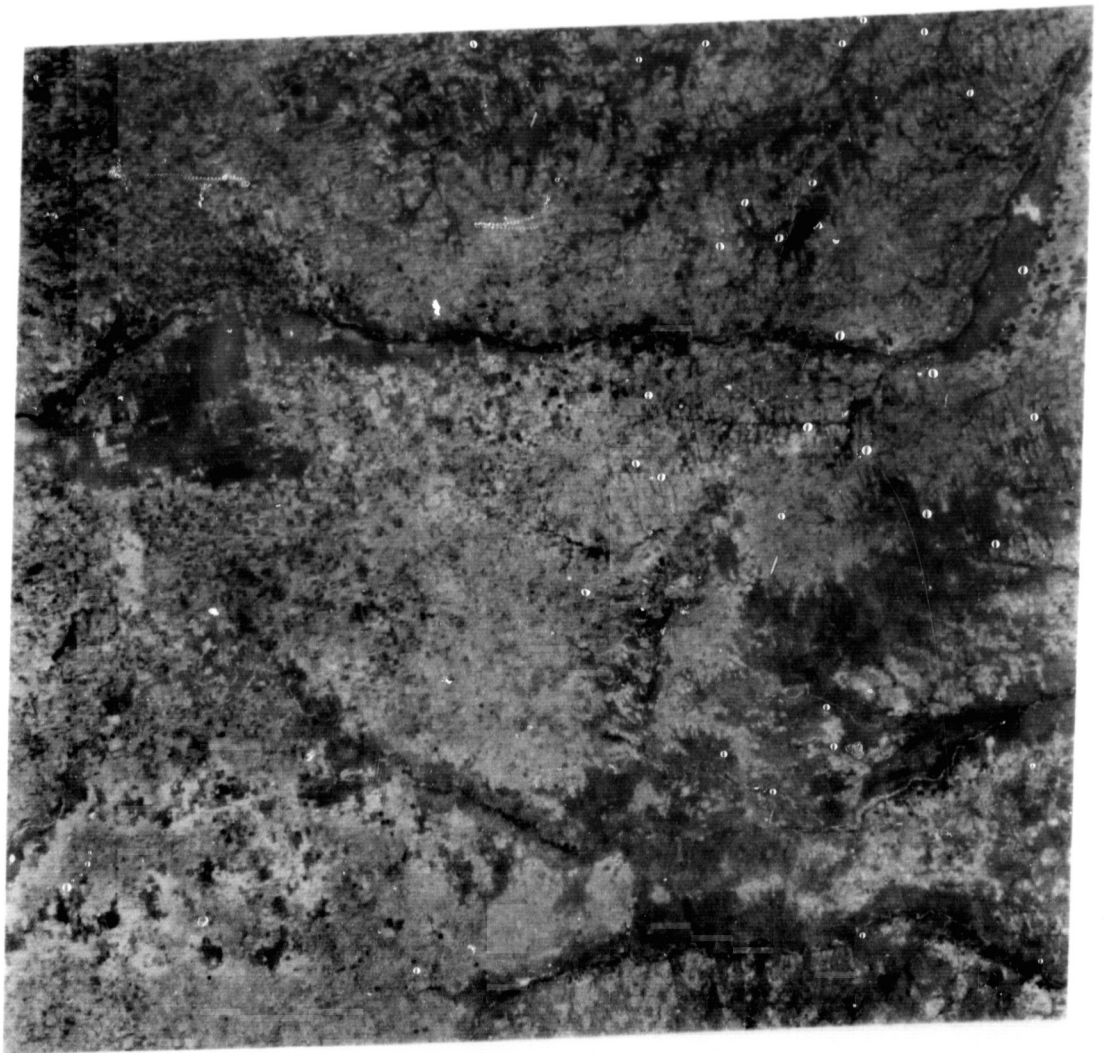


Figure 11. ERTS Image No. 1330-16515 (MSS Band 5) Over Garden City, Kansas, Taken on June 18, 1973.



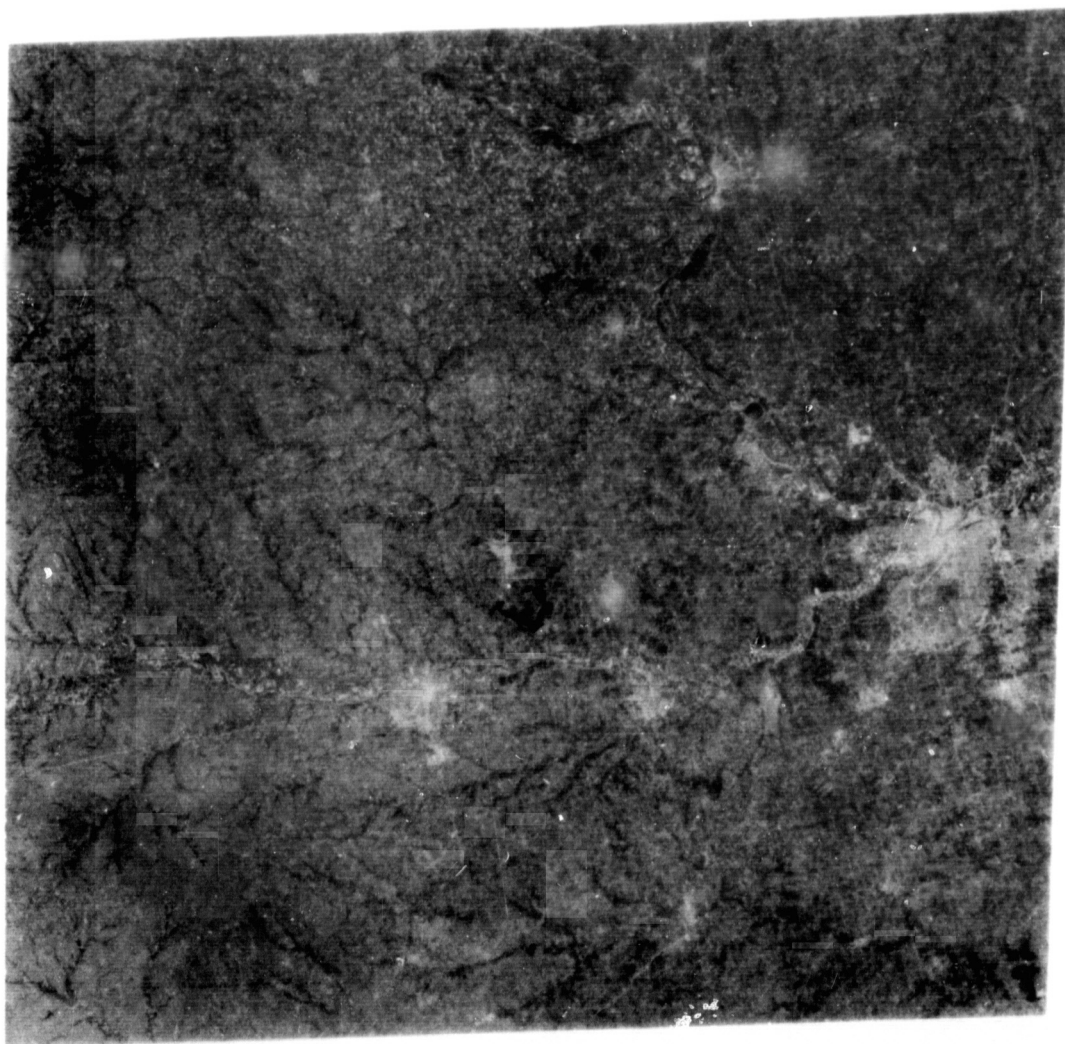


Figure 12. ERTS Image No. 1021-16333 Over (MSS Band 5) Kansas City,  
Taken on August 13, 1972.

	BAND 5	BAND 6	BAND 7	(BAND 5) <sup>2</sup>	(BAND 6) <sup>2</sup>	(BAND 7) <sup>2</sup>	(BAND 5) x(BAND 6)	(BAND 5) x(BAND 7)
BAND 5	1							
BAND 6	F(2)	1						
BAND 7	F(3)	F(9)	1					
(BAND 5) <sup>2</sup>	F(4)	F(10)	F(15)	1				
(BAND 6) <sup>2</sup>	F(5)	F(11)	F(16)	F(20)	1			
(BAND 7) <sup>2</sup>	F(6)	F(12)	F(17)	F(21)	F(24)	1		
(BAND 5)(BAND 6)	F(7)	F(13)	F(18)	F(22)	F(25)	F(27)	1	
(BAND 5)(BAND 7)	F(8)	F(14)	F(19)	F(23)	F(26)	F(28)	F(29)	1

ENTROPY = F(1)

FIGURE 13. DESIGNATION OF THE ENTROPY MEASURE AND ELEMENTS OF THE CORRELATION MATRIX AS FEATURE VECTOR COMPONENTS FOR CROSS-BAND TEXTURE ANALYSIS OF MULTI-IMAGES.

Table 12 shows the contingency table for all 17 single-image texture features over image 1021-16333 with an average correct classification of 86.2%. Table 13 shows the resulting contingency table over the same image for the 29 cross-band texture features with an increased correct classification of 88.8%. The spectral processing on the four MSS bands for the same image gave four spectral features—the mean grey tone over the subimage for each spectral band. The resulting contingency table using these four spectral features is shown in Table 14. The relatively good performance, 73.9% correct, shows a significant amount of land-use information is contained in the four spectral bands, accounting for the better performance of the cross-band texture analysis over the single-image texture analysis.

It is interesting to see the classification accuracy of the single-image texture greatly improve to 93.8% with the addition of the four spectral features to the original 17 texture features, as shown in Table 15. As expected, the addition of the four spectral features to the first 26 of the 29 cross-band texture features does not improve the classification accuracy as well as with the single-image texture. Table 16 shows the resulting contingency table with an average correct classification of 91.6%. Note the higher estimate for the standard deviation of the probability for correct classification with the textural plus spectral features.

These tables show that the single-image texture procedure does well in extracting texture information, but for this data set, the cross-band texture procedure by itself performs slightly better by extracting more information texturally and spectrally.

The remaining tables show the effect upon classification accuracy of using fewer features. In each case it can be seen that the fewer the number of features used, the lower the average correct classification. Also, it can be seen that the Bayes classifier performs slightly better than the piecewise linear classifier.

Tables 20 thru 24 show that as the number of components is increased, the spectral plus cross-band texture reaches a limiting accuracy of approximately 92%. Ground truth assignment errors could easily account for the remaining 8% error.

### VII.1 Summary of Classification Results

It is apparent from these classification experiments that both texture extraction procedures complement each other in that they extract different kinds of texture information. When used without spectral features, the cross-band texture procedure performed better. However, when the spectral features were added, the single-image texture plus spectral features performed better than the cross-band texture plus spectral features. This indicates that the cross-band texture procedure does well in

extracting more information texturally and spectrally than the single-image texture procedure. However, when the spectral information is made available to the single-image texture procedure, it performs better than the cross-band texture plus spectral features.

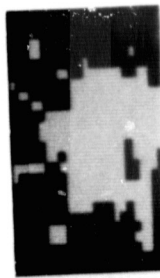
Programs used in these experiments can be found in Appendix IV. The single-image texture programs are under the Texture Analysis Programs with mainline MAINLN. And the cross-band texture programs are in the Cross-Band Texture Analysis Programs with mainline SPECTR. RCLASS is the mainline program for the piecewise linear classifier. Some of the cross-band texture feature components can be seen in Figure 14. Figure 14a shows the ground truth assignment for the image 1021-16333 (Figure 12) for the area near Kansas City. At the bottom of the image, grassland was inserted to give an idea of how well the features separate the three categories: cropland, urban area, and grassland.

## VIII. CONCLUSION

The procedure developed here for the extraction of texture information from ERTS multi-images gives encouraging results. The classification experiments show that the cross-band texture procedure can be used successfully in automatic land-use classification of multi-images over Kansas.

The cross-band texture procedure is a natural extension of the previous single-image texture extraction procedure based upon angular nearest neighbor grey tone spatial dependence matrices. It retains the power of the previous approach to texture by characterizing the spatial inter-relationships, or co-occurrences, of the grey tone N-tuples present in a texture pattern in such a way as to be invariant under linear grey tone transformations. And both procedures are simple to employ, economical, and require a minimum of core-storage (see Figure 15).

Both procedures complement each other by extracting different kinds of textural information with the cross-band texture procedure using the cross-band spectral information contained in ERTS multi-images. Results indicate that the cross-band texture procedure does well by extracting more information texturally and spectrally than the single-image texture procedure when used alone. However, when the spectral information is made available to the single-image texture procedure, it performs better than the cross-band texture plus spectral features in classifying texturally distinct land-use categories from ERTS multi-images over Kansas.



← Cropland

← Urban Area

Figure 14-a. Ground Truth Map

← Grassland

Figure 14-b. F(16)



Figure 14-e. F(29)



Figure 14-d. F(24)

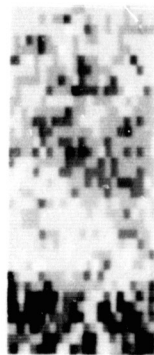


Figure 14-c. F(20)

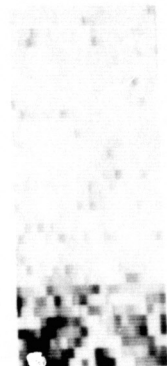


Figure 14. Examples of Cross-Band Texture Feature Vector Components; Resolution Cell Size is 32 x 32

	CORE REQUIRED	TOTAL PROCESSING TIME	TOTAL JOB COST	COST PER RESOLUTION CELL
SINGLE-IMAGE TEXTURE (64x64)	43k	0.788HR	\$223.	\$0.000252
CROSS-BAND TEXTURE (32x32)	25k	0.884HR	\$216.	\$0.000244

FIGURE 15. COMPARISON OF THE PERFORMANCE OF THE SINGLE-IMAGE TEXTURE ANALYSIS PROGRAMS ON MSS BAND 5 WITH THE CROSS-BAND TEXTURE ANALYSIS PROGRAMS ON MSS BANDS 4 THRU 7 IN PROCESSING THE SAME AREA, NEAR KANSAS CITY, ON ERTS IMAGE No. 1021-16333 USING A HONEYWELL 635 COMPUTER.



		ASSIGNED CATEGORY					ERRORS OF COMMISSION		
		COASTAL FOREST	ANNUAL GRASSLAND	URBAN AREA	WATER	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	COASTAL FOREST	5	0	5	0	10	5	50.	15.8
	ANNUAL GRASSLAND	0	7	0	0	7	0	0.	0.
	URBAN AREA	0	1	7	0	8	1	14.3	11.7
	WATER	0	0	0	5	5	0	0.	0.
	TOTAL	5	8	12	5	30		21.4	7.3
ERRORS OF COMMISSION	#ERRORS	0	1	7	0				
	%ERROR	0.	12.5	58.3	0.	17.7			

TABLE 1. CONTINGENCY TABLE FOR IMAGE NO. 1002-18134 USING MULTI-IMAGE TEXTURAL FEATURES, 6 COMPONENTS, AVERAGE CORRECT CLASSIFICATION ON TEST SET=80.0%\*

FEATURES	NO. OF SAMPLES IN TRAINING SET	NO. OF SAMPLES IN TEST SET	OVERALL ACCURACY OF TEST SET
MULTI-IMAGE TEXTURAL	34	30	80.0%
SINGLE-IMAGE TEXTURAL	260	172	70.5%

TABLE 2. RESULTS OF LAND-USE CLASSIFICATION EXPERIMENTS FROM ERTS IMAGE NO. 1002-18134 OVER MONTEREY BAY, CALIFORNIA.

\*UNLESS STATED OTHERWISE, ALL CONTINGENCY TABLES ARE DETERMINED USING THE PIECEWISE LINEAR CLASSIFICATION PROGRAMS (RCLASS).

		ASSIGNED CATEGORY					ERRORS OF COMMISSION		
		GRASSLAND	LARGE FIELDS	SMALL FIELDS	WATER	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	GRASSLAND	47	11	5	0	63	16	25.4	5.5
	LARGE FIELDS	3	142	32	1	178	36	20.2	3.0
	SMALL FIELDS	17	46	52	0	115	63	31.3	4.6
	WATER	0	1	6	1	8	2	25.0	11.7
	TOTAL	67	200	95	2	364		25.5	2.5
ERRORS OF COMMISSION	#ERRORS	20	58	43	1				
	%ERROR	29.9	29.0	45.3	50.	38.6			

TABLE 3. CONTINGENCY TABLE FOR IMAGE NO. 1330-16515 USING MULTI-IMAGE TEXTURAL FEATURES, DISTANCE 1, 6 COMPONENTS. AVERAGE CORRECT CLASSIFICATION = 66.5%

FEATURES	NO. OF SAMPLES IN TRAINING SET	NO. OF SAMPLES IN TEST SET	OVERALL ACCURACY OF TEST SET
MULTI-IMAGE TEXTURAL	548	364	66.5%
SINGLE-IMAGE TEXTURAL	140	88	76%

TABLE 4. RESULTS OF LAND-USE CLASSIFICATION EXPERIMENTS FOR ERTS IMAGE NO. 1330-16515 AT DISTANCE 1.

REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR



		ASSIGNED CATEGORY					ERRORS OF COMMISSION		
		GRASSLAND	LARGE FIELDS	SMALL FIELDS	WATER	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	GRASSLAND	32	9	22	0	63	31	49.2	6.3
	LARGE FIELDS	7	147	24	0	178	31	17.4	2.8
	SMALL FIELDS	18	44	53	0	115	62	53.9	4.6
	WATER	0	4	4	0	8	4	50.0	0.
	TOTAL	57	204	103	0	364		42.6	2.5
ERRORS OF COMMISSION	#ERRORS	25	57	50	0				
	%ERROR	43.9	27.9	48.5	0.	30.1			

TABLE 5. CONTINGENCY TABLE FOR IMAGE NO. 1330-16515 USING MULTI-IMAGE TEXTURAL FEATURES, DISTANCE 8, 6 COMPONENTS. AVERAGE CORRECT CLASSIFICATION = 63.7%

FEATURES	NO. OF SAMPLES IN TRAINING SET	NO. OF SAMPLES IN TEST SET	OVERALL ACCURACY OF TEST SET
MULTI-IMAGE TEXTURAL	548	364	63.7%
SINGLE-IMAGE TEXTURAL	140	88	76%

TABLE 6. RESULTS OF LAND-USE CLASSIFICATION EXPERIMENTS FOR ERTS IMAGE NO. 1330-16515 AT DISTANCE 8.

		ASSIGNED CATEGORY					ERRORS OF COMMISSION		
		GRASSLAND	LARGE FIELDS	SMALL FIELDS	WATER	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	GRASSLAND	52	7	4	0	63	11	17.5	4.8
	LARGE FIELDS	2	145	31	0	178	33	18.5	2.9
	SMALL FIELDS	12	42	61	0	115	54	47.0	4.7
	WATER	0	1	6	1	8	7	87.5	11.7
	TOTAL	66	195	102	1	364		42.6	2.4
ERRORS OF COMMISSION	#ERRORS	14	50	41	0				
	%ERROR	21.2	25.6	40.2	0.	21.8			

TABLE 7. CONTINGENCY TABLE FOR IMAGE NO. 1330-16515 USING MULTI-IMAGE TEXTURAL FEATURES, DISTANCES 1 AND 8, 12 COMPONENTS. AVERAGE CORRECT CLASSIFICATION = 71%

FEATURES	NO. OF SAMPLES IN TRAINING SET	NO. OF SAMPLES IN TEST SET	OVERALL ACCURACY OF TEST SET
MULTI-IMAGE TEXTURAL	548	364	71%
SINGLE-IMAGE TEXTURAL	140	88	73%

TABLE 8. RESULTS OF LAND-USE CLASSIFICATION EXPERIMENTS FOR ERTS IMAGE NO. 1330-16515 USING BOTH DISTANCES 1 AND 8.



REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR

		ASSIGNED CATEGORY					ERRORS OF COMMISSION		
		GRASSLAND	LARGE FIELDS	SMALL FIELDS	WATER	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	GRASSLAND	49	4	6	1	60	11	18.3	5.0
	LARGE FIELDS	6	150	27	1	184	34	18.5	2.9
	SMALL FIELDS	10	34	67	2	113	46	40.7	4.6
	WATER	0	4	0	4	8	4	31.9	17.7
	TOTAL	65	192	100	8	365		31.9	2.3
ERRORS OF COMMISSION	#ERRORS	16	42	33	4				
	%ERROR	24.6	21.9	33.0	50.0	32.4			

TABLE 9. CONTINGENCY TABLE FOR IMAGE NO. 1330-16515 USING MULTI-IMAGE TEXTURAL FEATURES, DISTANCE 1,6 COMPONENTS. AVERAGE CORRECT CLASSIFICATION = 74.0% USING BAYES.

		ASSIGNED CATEGORY					ERRORS OF COMMISSION		
		GRASSLAND	LARGE FIELDS	SMALL FIELDS	WATER	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	GRASSLAND	31	10	19	0	60	29	48.3	6.5
	LARGE FIELDS	5	140	38	1	184	44	23.9	3.1
	SMALL FIELDS	12	46	51	4	113	62	54.9	4.7
	WATER	1	4	2	1	8	7	87.5	11.7
	TOTAL	49	200	110	6	365		53.7	2.6
ERRORS OF COMMISSION	#ERRORS	18	50	59	5				
	%ERROR	36.7	25.0	53.6	83.3	49.7			

TABLE 10. CONTINGENCY TABLE FOR IMAGE NO. 1330-16515 USING MULTI-IMAGE TEXTURAL FEATURES, DISTANCE 8,6 COMPONENTS. AVERAGE CORRECT CLASSIFICATION = 61.1% USING BAYES.

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		GRASSLAND	LARGE FIELDS	SMALL FIELDS	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	GRASSLAND	35	3	0	38	3	7.9	4.4
	LARGE FIELDS	2	12	2	16	4	25.0	10.8
	SMALL FIELDS	0	2	14	16	2	12.5	8.3
	TOTAL	37	17	16	70		15.1	4.0
ERRORS OF COMMISSION	#ERRORS	2	5	2				
	%ERROR	5.4	29.4	12.5	15.8			

TABLE 11. CONTINGENCY TABLE FOR IMAGE NO. 1330-16515 USING 9 OF THE 29 CROSS-BAND TEXTURAL FEATURES, DISTANCE 1. #TRAIN=103, #TEST=70, AVERAGE CORRECT CLASSIFICATION = 87.1% USING THE BAYES CLASSIFIER. THE 9 FEATURE COMPONENTS USED WERE: 1,2,5,6,9,12,20,24,27.



		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	61	2	1	64	3	4.7	2.6
	URBAN	6	18	0	24	6	25.0	8.8
	GRASSLAND	6	1	21	28	7	25.0	8.2
	TOTAL	73	21	22	116		18.2	3.2
ERRORS OF COMMISSION	#ERRORS	12	3	1				
	%ERROR	16.4	14.2	4.5	11.7			

TABLE 12. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING ALL 17 SINGLE-BAND TEXTURE FEATURES AT DISTANCE 1. #TRAIN=178, #TEST=116, AVERAGE CORRECT CLASSIFICATION = 86.2%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	150	15	3	168	18	10.7	2.4
	URBAN	11	81	0	92	11	12.0	3.4
	GRASSLAND	11	2	103	116	13	11.2	2.9
	TOTAL	172	98	106	376		11.3	1.6
ERRORS OF COMMISSION	#ERRORS	22	17	3				
	%ERROR	12.8	17.3	2.8	11.0			

TABLE 13. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING ALL 29 CROSS-BAND TEXTURE FEATURES AT DISTANCE 1. #TRAIN = 569, #TEST = 376, AVERAGE CORRECT CLASSIFICATION = 88.8%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	138	9	22	169	37	21.9	3.0
	URBAN	8	53	30	91	38	41.8	5.2
	GRASSLAND	20	8	84	112	28	25.0	4.1
	TOTAL	166	70	136	372		29.6	2.3
ERRORS OF COMMISSION	#ERRORS	28	17	52				
	%ERROR	16.9	24.3	38.2	26.5			

TABLE 14. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING SPECTRAL FEATURES ONLY. #TRAIN=564, #TEST=372  
AVERAGE CORRECT CLASSIFICATION = 73.9%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	62	2	0	64	2	3.1	2.2
	URBAN	4	19	1	24	5	20.8	8.3
	GRASSLAND	0	0	24	24	0	0.0	0.0
	TOTAL	66	21	25	112		8.0	2.3
ERRORS OF COMMISSION	#ERRORS	4	2	1				
	%ERROR	6.1	9.5	4.0	6.5			

TABLE 15. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING ALL 17 SINGLE-IMAGE TEXTURE PLUS THE 4 SPECTRAL FEATURES, 21 COMPONENTS. #TRAIN=175, #TEST=112  
AVERAGE CORRECT CLASSIFICATION = 93.8%



		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	155	12	1	168	13	7.7	2.1
	URBAN	12	80	0	92	12	13.0	3.5
	GRASSLAND	4	2	103	109	6	5.5	2.2
	TOTAL	171	94	104	369		8.7	1.4
ERRORS OF COMMISSION	#ERRORS	16	12	1				
	%ERROR	9.4	12.8	1.0	7.7			

TABLE 16. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 26 CROSS-BAND TEXTURE PLUS THE 4 SPECTRAL FEATURES, 30 COMPONENTS. #TRAIN=567, #TEST=369  
AVERAGE CORRECT CLASSIFICATION = 91.6%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	60	1	3	64	4	6.2	3.0
	URBAN	8	16	0	24	8	33.3	9.6
	GRASSLAND	5	1	22	28	6	21.4	7.8
	TOTAL	73	18	25	116		20.3	3.4
ERRORS OF COMMISSION	#ERRORS	13	2	3				
	%ERROR	17.8	11.1	12.0	13.6			

TABLE 17. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 8 OF THE 17 SINGLE-IMAGE TEXTURE FEATURES. #TRAIN=178, #TEST=116, AVERAGE CORRECT CLASSIFICATION = 84.5%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	138	22	8	168	30	17.9	3.0
	URBAN	21	68	3	92	24	26.1	4.6
	GRASSLAND	20	5	91	116	25	21.6	3.8
	TOTAL	179	95	102	376		21.9	2.1
ERRORS OF COMMISSION	#ERRORS	41	27	11				
	%ERROR	22.9	28.4	10.8	20.7			

TABLE 18. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 9 OF THE 29 CROSSBAND TEXTURE FEATURES.  
 #TRAIN=569, #TEST=376, AVERAGE CORRECT CLASSIFICATION = 79.0%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	142	25	4	171	29	17.0	2.9
	URBAN	20	66	5	91	25	27.5	4.7
	GRASSLAND	12	5	95	112	17	15.2	3.4
	TOTAL	174	96	104	374		19.9	2.0
ERRORS OF COMMISSION	#ERRORS	32	30	9				
	%ERROR	18.4	31.3	8.7	19.5			

TABLE 19. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 9 OF THE 29 CROSS-BAND TEXTURE FEATURES.  
 #TRAIN=562, #TEST=374, AVERAGE CORRECT CLASSIFICATION = 81.0% USING THE BAYES CLASSIFIER.



		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	158	6	2	166	8	4.8	1.7
	URBAN	13	58	18	89	31	34.8	5.1
	GRASSLAND	9	10	93	112	19	17.0	3.5
	TOTAL	180	74	113	367		18.9	1.9
ERRORS OF COMMISSION	#ERRORS	22	16	20				
	%ERROR	12.2	21.6	17.7	17.2			

TABLE 20. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 6 OF THE 29 CROSS-BAND TEXTURE PLUS THE 4 SPECTRAL FEATURES, #TRAIN=569, #TEST=367, AVERAGE CORRECT CLASSIFICATION = 84.2%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	61	2	1	64	3	4.7	2.6
	URBAN	3	19	2	24	5	20.8	8.3
	GRASSLAND	0	0	24	24	0	0.0	0.0
	TOTAL	64	21	27	112		8.5	2.4
ERRORS OF COMMISSION	#ERRORS	3	2	3				
	%ERROR	4.7	9.5	11.1	8.4			

TABLE 21. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 9 OF THE 17 SINGLE-IMAGE TEXTURE PLUS THE 4 SPECTRAL FEATURES, #TRAIN=175, #TEST=112, AVERAGE CORRECT CLASSIFICATION = 92.9%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	161	7	0	168	7	4.2	1.5
	URBAN	15	77	0	92	15	16.3	3.9
	GRASSLAND	9	6	97	112	15	13.4	3.2
	TOTAL	185	90	97	372		11.3	1.6
ERRORS OF COMMISSION	#ERRORS	24	13	0				
	%ERROR	13.0	14.4	0.0	9.1			

TABLE 22. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 11 OF THE 29 CROSS-BAND TEXTURE PLUS THE 4 SPECTRAL FEATURES, #TRAIN=564, #TEST=372, AVERAGE CORRECT CLASSIFICATION = 90.1%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	159	8	0	167	8	4.8	1.7
	URBAN	15	76	1	92	16	17.4	4.0
	GRASSLAND	7	5	100	112	12	10.7	2.9
	TOTAL	181	89	101	371		11.0	1.5
ERRORS OF COMMISSION	#ERRORS	22	13	1				
	%ERROR	12.2	14.6	1.0	9.3			

TABLE 23. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 16 OF THE 29 CROSS-BAND TEXTURE PLUS THE 4 SPECTRAL FEATURES, #TRAIN=565, #TEST=371, AVERAGE CORRECT CLASSIFICATION = 90.3%

		ASSIGNED CATEGORY				ERRORS OF COMMISSION		
		CROPLAND	URBAN	GRASSLAND	TOTAL	#ERR	%ERR	%SD
TRUE CATEGORY	CROPLAND	156	9	1	166	10	6.0	1.8
	URBAN	13	79	0	92	13	14.1	3.6
	GRASSLAND	6	1	103	110	7	6.4	2.3
	TOTAL	175	89	104	368		8.8	1.4
ERRORS OF COMMISSION	#ERRORS	19	10	1				
	%ERROR	10.9	11.2	1.0	7.7			

TABLE 24. CONTINGENCY TABLE FOR ERTS IMAGE 1021-16333 USING FIRST 21 OF THE 29 CROSS-BAND TEXTURE PLUS THE 4 SPECTRAL FEATURES, #TRAIN=568, #TEST=368, AVERAGE CORRECT CLASSIFICATION = 91.9%

## APPENDIX I

### TEXTURAL FEATURES OBTAINED FROM THE GREY TONE DEPENDENCE MATRIX

In this appendix, we define 17 textural features which are computed for each of the four angular grey tone dependence matrices.

The following notation will be used in defining the 17 textural features.

$P(i, j)$  -  $(i, j)^{\text{th}}$  entry in a particular grey tone dependence matrix.

$\left. \begin{array}{l} \frac{P_x(i)}{\#R} \\ \frac{P_y(i)}{\#R} \end{array} \right\}$  -  $i^{\text{th}}$  entry in the marginal distributions of  $P(i, j)$  obtained by summing rows and columns of  $P(i, j)$  respectively.

$\#R$  - number of resolution cell pairs which were considered in computing the entries in  $P(i, j)$ .

$N_g$  - number of distinct grey tone values in the image.

$\mu$  - mean of  $P(i, j)/\#R$ .

$\frac{P_{x+y}}{\#R}(i)$  -  $i^{\text{th}}$  entry in the distribution of the sum of grey tones of neighboring resolution cells.

$\frac{P_{x-y}}{\#R}(i)$  -  $i^{\text{th}}$  entry in the distribution of the absolute differences in the grey tones of neighboring resolution cells.



## TEXTURAL FEATURES

### 1. Angular Second Moment:

$$f_1 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} \left\{ \frac{P(i,j)}{\#R} \right\}^2$$

### 2. Entropy:

$$f_2 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} - \left( \frac{P(i,j)}{\#R} \right) \log \left( \frac{P(i,j)}{\#R} \right)$$

### 3. Correlation:

$$f_3 = \frac{\sum_{i=1}^{N_g} \sum_{j=1}^{N_g} ij \left\{ \frac{P(i,j)}{\#R} \right\} - \mu_x \mu_y}{\sigma_x \sigma_y}$$

where  $\mu_x$  and  $\sigma_x$  are the mean and standard deviation of  $P_x$ , and  $\mu_y$  and  $\sigma_y$  are the mean and standard deviation of  $P_y$ .

### 4. Sum of Squares on x:

$$f_4 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} (i-\mu)^2 \left\{ \frac{P(i,j)}{\#R} \right\}$$

### 5. Product Moment:

$$f_5 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} (i-\mu)(j-\mu) \left\{ \frac{P(i,j)}{\#R} \right\}$$

6. Inverse Moment:

$$f_6 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} \frac{1}{1+(i-j)^2} \left\{ \frac{P(i,j)}{\#R} \right\}$$

7. Difference Moment:

$$f_7 = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} (i-j)^2 \left\{ \frac{P(i,j)}{\#R} \right\}$$

8. Sum Average:

$$f_8 = \sum_{i=1}^{2N_g} i \left\{ \frac{P_{x+y}(i)}{\#R} \right\}$$

9. Mean:

$$f_9 = \frac{1}{N_g} \sum_{i=1}^{N_g} \left\{ \frac{P(i,j)}{\#R} \right\}$$

10. Sum Variance:

$$f_{10} = \text{variance of } P_{x+y} / \#R$$

11. Sum Entropy:

$$f_{11} = \sum_{i=1}^{2N_g} - \left\{ \frac{P_{x+y}(i)}{\#R} \right\} \log \left\{ \frac{P_{x+y}(i)}{\#R} \right\}$$

12. Contrast:

$$f_{12} = \sum_{n=0}^{N_g-1} n^2 \left( \sum_{\substack{i=1 \\ |i-j|=n}}^{N_g} \sum_{j=1}^{N_g} \left\{ \frac{P(i,j)}{\#R} \right\} \right)$$

13. Difference Variance:

$$f_{13} = \text{variance of } \frac{P_{x-y}}{\#R}$$

14. Difference Entropy:

$$f_{14} = \sum_{i=0}^{N_g-1} - \left\{ \frac{P_{x-y}(i)}{\#R} \right\} \log \left\{ \frac{P_{x-y}(i)}{\#R} \right\}$$

15, 16, 17. Additional Measures of Correlation:

$$f_{15} = \frac{HXY - HXY1}{\max HX, HY}$$

$$f_{16} = \sqrt{1 - \exp[-2.0(HXY2 - HXY)]}$$

$$f_{17} = \sqrt{\text{second largest eigenvalue of } QQ^T}^*$$

where HX and HY are the entropies of the marginals of the transition matrix before quantization, HXY is the entropy of the transition matrix, and HXY2 is the entropy of the product distribution of the marginals before quantization;

$$Q(i, j) = P(i, j) / \sqrt{P_x(i)P_y(j)}$$

\*  $f_{17}$  is the maximal correlation coefficient.

## APPENDIX II

### N-DIMENSIONAL SPHERICAL COORDINATE SYSTEMS AND ELLIPSOIDALLY SYMMETRIC DISTRIBUTIONS

We illustrate the N-dimensional spherical coordinate system in the calculation of the volume of the N-dimensional hypersphere. Next we show how suitable functions can be used to define ellipsoidally symmetric density functions and we determine the normalizing constant for any function. Finally, we show that for any ellipsoidally symmetric density  $f(\sqrt{x^T Ax})$ , the matrix A is proportional to the inverse covariance matrix of x and we determine the constant of proportionality.

#### II.1 Volume of an N-dimensional Hypersphere

Let V be the volume of a N-dimensional hypersphere of radius  $r_0$ . By definition

$$v = \int \int \dots \int dx_1 dx_2 \dots dx_N$$

$$\sqrt{\sum_{i=1}^N x_i^2} \leq r_0$$

To evaluate this N-fold integral, we transform to spherical coordinates.

$$x_1 = r \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-3} \cos \theta_{N-2} \cos \theta_{N-1}$$

$$x_2 = r \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-3} \cos \theta_{N-2} \sin \theta_{N-1}$$

$$x_3 = r \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-3} \sin \theta_{N-2}$$

$$\vdots$$

$$x_j = r \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-j} \sin \theta_{N-j}$$

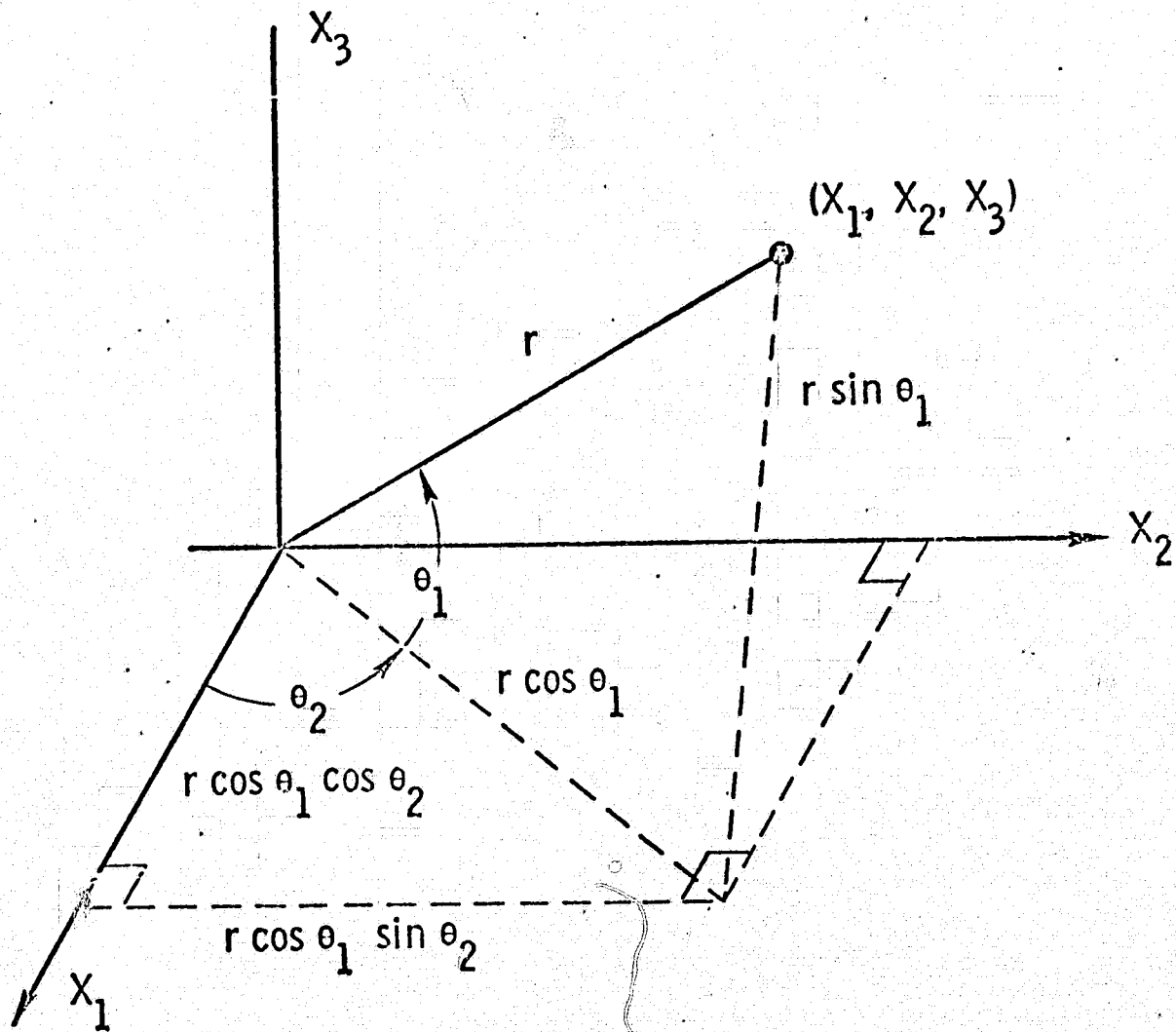
$$\vdots$$

$$x_N = r \sin \theta_1$$

Figure 16 illustrates the geometry of the spherical coordinate system we use for a 3-dimensional system.



### Three-Dimensional Spherical Coordinate System



$$\left. \begin{aligned} x_1 &= r \cos \theta_1 \cos \theta_2 \\ x_2 &= r \cos \theta_1 \sin \theta_2 \\ x_3 &= r \sin \theta_1 \end{aligned} \right\}$$

Transformation between rectangular coordinate system and spherical coordinate system.

Figure 16 Three-Dimensional Spherical Coordinate System

The Jacobian  $J$  of this transformation is defined by the determinant  $J$ .

$$J = \begin{vmatrix} \frac{\partial x_1}{\partial r} & \frac{\partial x_2}{\partial r} & \dots & \frac{\partial x_N}{\partial r} \\ \frac{\partial x_1}{\partial \theta_1} & \frac{\partial x_2}{\partial \theta_1} & \dots & \frac{\partial x_N}{\partial \theta_1} \\ \vdots & \vdots & \dots & \vdots \\ \frac{\partial x_1}{\partial \theta_{N-1}} & \frac{\partial x_2}{\partial \theta_{N-1}} & \dots & \frac{\partial x_N}{\partial \theta_{N-1}} \end{vmatrix}$$

$$= \begin{vmatrix} \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-1} & \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-2} \sin \theta_{N-1} & \dots & \sin \theta_1 \\ -r \sin \theta_1 \cos \theta_2 \dots \cos \theta_{N-1} & -r \sin \theta_1 \cos \theta_2 \dots \cos \theta_{N-2} \sin \theta_{N-1} & \dots & r \cos \theta_1 \\ -r \cos \theta_1 \sin \theta_2 \dots \cos \theta_{N-1} & -r \cos \theta_1 \sin \theta_2 \dots \cos \theta_{N-2} \sin \theta_{N-1} & \dots & 0 \\ \vdots & \vdots & \dots & \vdots \\ -r \cos \theta_1 \cos \theta_2 \dots \sin \theta_{N-1} & r \cos \theta_1 \cos \theta_2 \dots \cos \theta_{N-2} \cos \theta_{N-1} & \dots & 0 \end{vmatrix}$$

To find the value of the Jacobian, factor  $r$  out of the last  $(N-1)$  rows and from each column factor out its first entry.

$$J = r^{N-1} \cos^{N-1} \theta_1 \cos^{N-2} \theta_2 \dots \cos \theta_{N-1} \sin \theta_1 \sin \theta_2 \dots \sin \theta_{N-1}$$

$$\begin{vmatrix} 1 & 1 & 1 & \dots & 0 \\ -\tan \theta_1 & -\tan \theta_1 & -\tan \theta_1 & \dots & \cot \theta_1 \\ -\tan \theta_2 & -\tan \theta_2 & -\tan \theta_2 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & -\tan \theta_{N-3} & \dots & 0 \\ \vdots & -\tan \theta_{N-2} & \cot \theta_{N-2} & \dots & 0 \\ -\tan \theta_{N-1} & \cot \theta_{N-1} & 0 & \dots & 0 \end{vmatrix}$$

Subtracting column 2 from column 1, column 3 from column 2, .....  
column N from column N-1 there results

$$\begin{vmatrix} 0 & 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & 0 & \dots & -\tan \theta_1 & -\cot \theta_1 & \cot \theta_1 \\ \vdots & \vdots & \vdots & \ddots & \cot \theta_2 & & 0 \\ \vdots & \vdots & \vdots & \ddots & 0 & & \vdots \\ \vdots & \vdots & -\tan \theta_{N-3} & -\cot \theta_{N-3} & \vdots & & \vdots \\ \vdots & -\tan \theta_{N-2} & -\cot \theta_{N-2} & \cot \theta_{N-2} & \vdots & & \vdots \\ -\tan \theta_{N-1} & -\cot \theta_{N-1} & \cot \theta_{N-1} & 0 & \dots & 0 & 0 \end{vmatrix}$$

Since all entries in the upper left triangle are zero, the value of the determinant is easily found as minus one times the product of entries on the lower left to upper right diagonal.

$$J = r^{N-1} \cos^{N-1} \theta_1 \cos^{N-2} \theta_2 \dots \cos \theta_{N-1} \sin \theta_1 \sin \theta_2 \dots \sin \theta_{N-1} (-1)^{N-1} \pi^{N-1} (-\tan \theta_1 \dots \cot \theta_{N-1})$$

Notice that  $\tan \theta + \cot \theta = \frac{1}{\sin \theta \cos \theta}$ . Now upon simplifying we obtain

$$J = (-1)^{N-1} r^{N-1} \cos^{N-2} \theta_1 \cos^{N-3} \theta_2 \dots \cos \theta_{N-2}$$

and  $|J| = r^{N-1} \cos^{N-2} \theta_1 \cos^{N-3} \theta_2 \dots \cos \theta_{N-2}$  since

$$\cos \theta_i > 0 \text{ for } -\pi/2 \leq \theta_i \leq \pi/2, i=1, 2, \dots, N-2.$$

In spherical coordinates the volume  $V$  of the  $N$ -dimensional hypersphere of radius  $r_0$  is readily evaluated.

$$V = \int_{r=0}^{r_0} \int_{\theta_1=-\pi/2}^{\pi/2} \dots \int_{\theta_{N-2}=-\pi/2}^{\pi/2} \int_{\theta_{N-1}=0}^{2\pi} r^{N-1} \cos^{N-2} \theta_1 \cos^{N-3} \theta_2 \dots \cos \theta_{N-2} dr d\theta_1 \dots d\theta_{N-1}$$

Separating the integrations,

$$V = \int_{r=0}^{r_0} r^{N-1} dr \int_{\theta_1=-\pi/2}^{\pi/2} \cos^{N-2} \theta_1 d\theta_1 \dots \int_{\theta_{N-2}=-\pi/2}^{\pi/2} \cos \theta_{N-2} d\theta_{N-2} \int_{\theta_{N-1}=0}^{2\pi} d\theta_{N-1}$$

Since  $\int_{\theta=-\pi/2}^{\pi/2} \cos^N \theta d\theta = \frac{\Gamma\left(\frac{N+1}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N+2}{2}\right)}$ ,

$$\begin{aligned}
 V &= \frac{r_0^N}{Z} \left[ \frac{\Gamma\left(\frac{N-1}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N}{2}\right)} \right] \left[ \frac{\Gamma\left(\frac{N-2}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N-1}{2}\right)} \right] \dots \left[ \frac{\Gamma\left(\frac{2}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{3}{2}\right)} \right]^{2\pi} \\
 &= \frac{r_0^N}{Z} 2\pi \frac{\Gamma\left(\frac{1}{2}\right)^{N-2} \Gamma(1)}{\Gamma\left(\frac{N}{2}\right)}
 \end{aligned}$$

But  $\Gamma\left(\frac{1}{2}\right) = \pi^{1/2}$  and  $\Gamma(1) = 1$

$$V = \frac{r_0^N}{Z} \frac{2\pi \pi^{\frac{N-2}{2}}}{\Gamma\left(\frac{N}{2}\right)} = \frac{2 r_0^N}{Z} \frac{\pi^{\frac{1}{2}N}}{\Gamma\left(\frac{N}{2}\right)}$$

## II.2 Suitable Functions for Ellipsoidally Symmetric Distribution.

Suppose  $f$  is a real function, defined on domain  $R$ , a subset of  $[0, \infty]$ , and satisfying  $f(\mu) \geq 0$  for all  $\mu$  in  $R$  and  $\int \mu^k f(\mu) d\mu$  is finite for  $k \leq N+1$ . We show that  $f$  is suitable for defining an ellipsoidally symmetric density function and we determine the constant  $c$  so that  $f(\sqrt{x^T A x})$  is an ellipsoidally symmetric density.

Let  $A$  be a  $N \times N$  symmetric positive definite matrix and  $X$  an  $N \times 1$  vector. Consider the ellipsoidally symmetric function  $f(\sqrt{x^T A x})$ . We wish to determine a constant  $C$  such that  $f(\sqrt{x^T A x})$  is a density function.

$$\text{It is clear that } C = \frac{1}{\int_{\sqrt{x^T A x} \in R} \dots \int f(\sqrt{x^T A x}) dx_1 \dots dx_N}.$$

To determine the value of the integral, we will make a transformation which rotates and scales. Let  $T$  be an orthonormal matrix such that  $T^T A T = D$ , where  $D$  is a diagonal matrix. Make the change of variables

$$X = T D^{-1/2} z.$$

The Jacobian  $J$  of this transformation is

$$J = \begin{vmatrix} \frac{\partial x_1}{\partial z_1} & \frac{\partial x_2}{\partial z_1} & \dots & \frac{\partial x_N}{\partial z_1} \\ \frac{\partial x_1}{\partial z_2} & \frac{\partial x_2}{\partial z_2} & \dots & \frac{\partial x_N}{\partial z_2} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial x_1}{\partial z_N} & \frac{\partial x_2}{\partial z_N} & \dots & \frac{\partial x_N}{\partial z_N} \end{vmatrix} = |T D^{-1/2}| = |T| |D|^{-1/2}$$

Since  $T$  is an orthonormal matrix,  $|T| = 1$  and

$$|D| = |T'AT| = |T'| |A| |T| = |A|.$$

So the Jacobian is the determinant  $|A|^{-1/2}$  which is positive since  $A$  is positive definite.

$$I = \int \dots \int f(\sqrt{x'Ax}) dx_1 \dots dx_N$$

$$\sqrt{x'Ax} \in \mathbb{R}$$

$$= |A|^{-1/2} \int \dots \int f(\sqrt{z'D^{-1/2}T'ATD^{-1/2}z}) dz_1 \dots dz_N$$

$$\sqrt{z'D^{-1/2}T'ATD^{-1/2}z} \in \mathbb{R}$$

$$= |A|^{-1/2} \int \dots \int f(z'z) dz_1 \dots dz_N$$

$$\sqrt{z'z} \in \mathbb{R}$$

Now change to spherical coordinates.

$$z_1 = r \cos\theta_1 \cos\theta_2 \dots \cos\theta_{N-1}$$

$$z_2 = r \cos\theta_1 \cos\theta_2 \dots \cos\theta_{N-2} \sin\theta_{N-1}$$

⋮

⋮

$$z_j = r \cos\theta_1 \dots \cos\theta_{N-j} \sin\theta_{N-j+1}$$

⋮

⋮

$$z_N = r \sin\theta_1$$

The Jacobian of this transformation is  $(-1)^{N-1} r^{N-1} \cos^{N-2} \theta_1 \cos^{N-2} \theta_2 \dots \cos \theta_{N-2}$ .

$$I = |A|^{-1/2} \int_{r \in R} \int_{\theta_1 = -\pi/2}^{\pi/2} \dots \int_{\theta_{N-1} = 0}^{2\pi} f(r) r^{N-1} \cos^{N-2} \theta_1 \cos^{N-3} \theta_2 \dots \cos \theta_{N-2} dr d\theta_1 \dots d\theta_{N-1}$$

$$= |A|^{-1/2} \int_{r \in R} r^{N-1} f(r) dr \int_{\theta_1 = -\pi/2}^{\pi/2} \cos^{N-2} \theta_1 d\theta_1 \int_{\theta_2 = -\pi/2}^{\pi/2} \cos^{N-3} \theta_2 d\theta_2 \dots$$

$$\int_{\theta_{N-2} = -\pi/2}^{\pi/2} \cos \theta_{N-2} d\theta_{N-2} \int_{\theta_{N-1} = 0}^{2\pi} d\theta_{N-1}$$

Since  $\int_{\theta = -\pi/2}^{\pi/2} \cos^N \theta d\theta = \frac{\Gamma\left(\frac{N+1}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N+2}{2}\right)}$ , the integrals are readily evaluated.



$$I = |A|^{-1/2} \int_{r \in R} r^{N-1} f(r) dr \left[ \frac{\Gamma\left(\frac{N-1}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N}{2}\right)} \frac{\Gamma\left(\frac{N-2}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N-1}{2}\right)} \dots \frac{\Gamma\left(\frac{2}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{3}{2}\right)} \right] 2\pi$$

$$= |A|^{-1/2} \int_{r \in R} r^{N-1} f(r) dr \frac{\Gamma\left(\frac{1}{2}\right)^{N-2}}{\Gamma\left(\frac{N}{2}\right)} 2\pi$$

$$= \frac{2(\pi)^{N/2}}{|A|^{1/2} \Gamma\left(\frac{N}{2}\right)} \int_{r \in R} r^{N-1} f(r) dr$$

Therefore, the constant  $c$  is

$$c = \frac{|A|^{1/2} \Gamma\left(\frac{N}{2}\right)}{2(\pi)^{N/2}} \int_{r \in R} r^{N-1} f(r) dr$$

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Next we determine the normalizing constant  $c$  for the forms  $e^{-\frac{1}{2}u^2}$  and  $(1+u^2)^{-m}$ .

### Case 1. Multivariate Normal

The density function for the multivariate normal distribution is of the form

$$f(\sqrt{x'Ax}) = e^{-\frac{1}{2}x'Ax}, \quad 0 \leq x'Ax \leq \infty$$

$$\text{or, } f(r) = e^{-\frac{1}{2}r^2}, \quad 0 \leq r \leq \infty.$$

Since  $\int_{r \in R} r^{N-1} f(r) dr = \int_0^\infty r^{N-1} e^{-\frac{1}{2}r^2} dr = \int_0^\infty (2u)^{\frac{N-2}{2}} e^{-u} du = 2^{\frac{N-2}{2}} \Gamma\left(\frac{N}{2}\right)$

then the normalizing constant is

$$c = \frac{\Gamma\left(\frac{N}{2}\right)}{2\pi^{\frac{N}{2}} |A|^{-\frac{1}{2}} \int_{r \in R} r^{N-1} f(r) dr} = \frac{\Gamma\left(\frac{N}{2}\right)}{2\pi^{\frac{N}{2}} |A|^{-\frac{1}{2}} 2^{\frac{N-2}{2}} \Gamma\left(\frac{N}{2}\right)}$$

$$= \frac{1}{(2\pi)^{\frac{N}{2}} |A|^{\frac{1}{2}}}$$

and,  $f(\sqrt{x'Ax}) = \frac{1}{(2\pi)^{\frac{N}{2}} |A|^{\frac{1}{2}}} e^{-\frac{1}{2}x'Ax}, \quad 0 \leq x'Ax \leq \infty.$

### Case 2. Multivariate Pearson Type VII.

Let  $f(\sqrt{x'Ax}) = (1+x'Ax)^{-m}, \quad 0 \leq x'Ax \leq \infty,$

then,

$$\int_{r \in R} r^{N-1} f(r) dr = \int_0^\infty r^{N-1} (1+r^2)^{-m} dr = \int_0^1 \left(\frac{1-u}{u}\right)^{\frac{N-2}{2}} u^m \frac{du}{2u^2}$$

$$= \frac{1}{2} \frac{\Gamma\left(\frac{N}{2}\right) \Gamma\left(m - \frac{N}{2}\right)}{\Gamma(m)}, \quad m > \frac{N}{2}.$$

And the normalizing constant is

$$c = \frac{\Gamma\left(\frac{N}{2}\right) \Gamma(m)}{2\pi^{\frac{N}{2}} |A|^{-\frac{1}{2}} \frac{1}{2} \Gamma\left(\frac{N}{2}\right) \Gamma\left(m - \frac{N}{2}\right)} = \frac{\Gamma(m)}{\pi^{\frac{N}{2}} |A|^{-\frac{1}{2}} \Gamma\left(m - \frac{N}{2}\right)}$$

and,  $f(\sqrt{x'Ax}) = \frac{\Gamma(m) |A|^{\frac{1}{2}}}{\pi^{\frac{N}{2}} \Gamma\left(m - \frac{N}{2}\right)} (1+x'Ax)^{-m}, \quad m > \frac{N}{2}, \quad 0 \leq x'Ax \leq \infty.$

## II.3 Covariance Matrix For Multivariate Distributions.

Given the density function  $f(\sqrt{x'Ax})$ , we want to find the covariance matrix  $\ddagger$ ,

$$\ddagger = E(xx') = c \int \dots \int_{\sqrt{X'AX} \in R} xx' f(\sqrt{x'Ax}) dx_1 \dots dx_N$$

where  $c$  is a normalizing constant and  $N$  is the dimension of  $x$ . Using the orthonormal transformation  $T^T A T = D$ , where  $D$  is a diagonal matrix, and scaling with  $x = T D^{-\frac{1}{2}} z$ ,

we have 
$$\mathcal{I} = c \int_{\sqrt{z^T z} \in R} \dots \int (T D^{-\frac{1}{2}} z)^T (z^T D^{-\frac{1}{2}} T^T) f(\sqrt{z^T z}) |A|^{-\frac{1}{2}} dz_1 \dots dz_N$$

since

$$x^T A x = z^T D^{-\frac{1}{2}} T^T A T D^{-\frac{1}{2}} z = z^T D^{-\frac{1}{2}} D D^{-\frac{1}{2}} z = z^T z$$

and

$$J = \begin{vmatrix} \frac{\partial x_1}{\partial z_1} & \dots & \frac{\partial x_N}{\partial z_1} \\ \frac{\partial x_1}{\partial z_2} & \dots & \frac{\partial x_N}{\partial z_2} \\ \dots & \dots & \dots \\ \frac{\partial x_1}{\partial z_N} & \dots & \frac{\partial x_N}{\partial z_N} \end{vmatrix} = |T D^{-\frac{1}{2}}| = |T| |D|^{-\frac{1}{2}} = |D|^{-\frac{1}{2}} = |T^T A T|^{-\frac{1}{2}} = |A|^{-\frac{1}{2}}$$

Rearranging, 
$$\mathcal{I} = c |A|^{-\frac{1}{2}} T D^{-\frac{1}{2}} \int \dots \int z^T z f(\sqrt{z^T z}) dz_1 \dots dz_N D^{-\frac{1}{2}} T^T,$$
  

$$\sqrt{z^T z} \in R$$

where  $z^T z$  is an  $N \times N$  matrix. Looking at the off diagonal terms, for  $i \neq j$ ,

$$\int \dots \int z_i z_j f(\sqrt{z_i z_j}) dz_1 \dots dz_N = 0 \quad \text{since we are integrating an odd function over even limits.}$$

function over even limits.

For terms of  $\mathcal{I}$  along the diagonal, for  $i=j$ ,

$$\int \dots \int z_i^2 f(\sqrt{z^T z}) dz_1 \dots dz_N = \int \dots \int z_i^2 f(\sqrt{z^T z}) dz_1 \dots dz_N$$

and changing to spherical coordinates,

$$= \int_{r \in R} \int_{\theta_1 = -\pi/2}^{\pi/2} \dots \int_{\theta_{N-2} = -\pi/2}^{\pi/2} \int_{\theta_{N-1} = 0}^{2\pi} r^2 \cos^2 \theta_1 \dots \cos^2 \theta_{N-1} f(r) r^{N-1} \cos^{N-2} \theta_1 \dots \cos \theta_{N-2} dr d\theta_1 \dots d\theta_{N-1}$$

$$= \int_{r \in R} r^{N+1} f(r) dr \int_{\theta_1 = -\pi/2}^{\pi/2} \cos^N \theta_1 d\theta_1 \dots \int_{\theta_{N-2} = -\pi/2}^{\pi/2} \cos^3 \theta_{N-2} d\theta_{N-2} \int_{\theta_{N-1} = -\pi/2}^{\pi/2} \cos^2 \theta_{N-1} d\theta_{N-1}$$

Since 
$$\int_{-\pi/2}^{\pi/2} \cos^k \theta d\theta = \frac{\Gamma\left(\frac{k+1}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{k+2}{2}\right)}$$
,

then

$$\int_{\sqrt{z^T z} \in R} \dots \int_{z_i^2} f(\sqrt{z^T z}) dz_1 \dots dz_N = \int_{r \in R} r^{N+1} f(r) dr \frac{\Gamma\left(\frac{N+1}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N+2}{2}\right)} \frac{\Gamma\left(\frac{N}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{N+1}{2}\right)} \dots \frac{\Gamma\left(\frac{4}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{5}{2}\right)} 2 \frac{\Gamma\left(\frac{3}{2}\right) \Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{4}{2}\right)}$$

$$= \frac{2\pi^{\frac{N}{2}}}{N \Gamma\left(\frac{N}{2}\right)} \int_{r \in R} r^{N+1} f(r) dr.$$

From II.2,

$$c = \frac{\Gamma\left(\frac{N}{2}\right) |A|^{\frac{1}{2}}}{2\pi^{\frac{N}{2}} \int_{r \in R} r^{N-1} f(r) dr}$$

so that

$$\Sigma = \frac{\Gamma\left(\frac{N}{2}\right) |A|^{\frac{1}{2}}}{2\pi^{\frac{N}{2}} \int_{r \in R} r^{N-1} f(r) dr} |A|^{-\frac{1}{2}} T D^{-\frac{1}{2}} \frac{2\pi^{\frac{N}{2}}}{N \Gamma\left(\frac{N}{2}\right)} \int_{r \in R} r^{N+1} f(r) dr D^{-\frac{1}{2}} T'$$

$$\text{and, } \Sigma = T D^{-1} T' \frac{\int_{r \in R} r^{N+1} f(r) dr}{N \int_{r \in R} r^{N-1} f(r) dr}$$

where  $T' A T = D$ , or  $D^{-1} = T' A^{-1} T$ .

So that

$$\Sigma = A^{-1} \frac{\int_{r \in R} r^{N+1} f(r) dr}{N \int_{r \in R} r^{N-1} f(r) dr}$$

Since the integrals are constants for any  $f$ , the covariance matrix is directly proportional to  $A^{-1}$ . We determine the constant of proportionality for the multivariate normal and Pearson Type VII distributions.

Case 1. Multivariate Normal

For the multivariate normal, the density is of the form  $f(r) = ce^{-\frac{1}{2}r^2}$  where  $r^2 = x'Ax$ ,  $0 \leq r \leq \infty$ .

$$\text{Since, } \int_0^\infty r^k e^{-\frac{1}{2}r^2} dr = 2 \cdot \frac{k-1}{2} \Gamma\left(\frac{k+1}{2}\right)$$

$$\text{then } \int_0^\infty r^{N+1} f(r) dr = c 2^{\frac{N}{2}} \Gamma\left(\frac{N+2}{2}\right)$$

$$\text{and } \int_0^\infty r^{N-1} f(r) dr = c 2^{\frac{N-2}{2}} \Gamma\left(\frac{N}{2}\right)$$

$$\text{so that } \Phi = A^{-1} \frac{2^{\frac{N}{2}} \frac{N}{2} \Gamma\left(\frac{N}{2}\right)}{N 2^{\frac{N}{2}-1} \Gamma\left(\frac{N}{2}\right)}$$

$$\text{and thus, } \Phi = A^{-1}$$

$$\text{or, } A = \Phi^{-1}$$

Case 2 Multivariate Pearson Type VII.

$f(r) = c(1+r^2)^{-m}$ ,  $0 \leq r \leq \infty$ .

$$\text{Since } \int_0^\infty r^{N+1} f(r) dr = \frac{c}{2} \frac{\Gamma\left(\frac{N}{2} + 1\right) \Gamma\left(m - \frac{N}{2} - 1\right)}{\Gamma(m)}$$

$$\text{and } \int_0^\infty r^{N-1} f(r) dr = \frac{c}{2} \frac{\Gamma\left(\frac{N}{2}\right) \Gamma\left(m - \frac{N}{2}\right)}{\Gamma(m)}$$

$$\text{then, } \Phi = A^{-1} \frac{\frac{cN}{2} \Gamma\left(\frac{N}{2}\right) \Gamma\left(m - \frac{N}{2} - 1\right)}{2N c \Gamma(m)} \frac{2 \Gamma(m)}{c \Gamma\left(\frac{N}{2}\right) \left(m - \frac{N}{2} - 1\right) \Gamma\left(m - \frac{N}{2} - 1\right)}$$

$$= A^{-1} \frac{1}{2 \left(m - \frac{N}{2} - 1\right)}$$

### APPENDIX III

#### NORMALIZATION PROCEDURE TO MAKE COVARIANCE MATRIX INVARIANT UNDER TRANSLATING AND SCALING TRANSFORMATIONS

Let  $\hat{\Sigma}_x$  be a covariance matrix for the difference vectors of grey tone N-tuples in a specified spatial relationship within a subimage. We transform the covariance matrix to obtain the normalized covariance matrix  $\hat{\Sigma}_y$  using  $y = Dx$ , where  $x$  is the difference vector and  $D$  is diagonal. Thus, assuming zero mean,

$$\begin{aligned}\hat{\Sigma}_y &= E(yy') = E(Dxx'D') \\ &= D E(xx') D' \\ &= D \hat{\Sigma}_x D \quad \text{since } D' = D.\end{aligned}$$

For normalization, we have

$$d_{ii} = \frac{1}{\sqrt{\sigma_{ii}}}$$

where  $\sigma_{ii}$  is the  $ii^{\text{th}}$  element of  $\hat{\Sigma}_x$  and is the variance  $\sigma_i^2$  of the  $i^{\text{th}}$  component of  $x$ .

Assume that all grey tone N-tuples have a scale factor  $a$  and an additive factor  $c$  so that for N-tuples  $ax_1 + c$  and  $ax_2 + c$ , the difference becomes

$$\begin{aligned}y &= (ax_1 + c) - (ax_2 + c) \\ y &= a(x_1 - x_2).\end{aligned}$$

Hence, translational effects due to bias terms are cancelled but scaling effects are marked by the diagonal transformation  $y = Ax$  so that the elements of the covariance matrix become

$$\begin{aligned}\hat{\Sigma} &= E(yy') = E(Axx'A') \\ \hat{\Sigma} &= A \hat{\Sigma}_x A\end{aligned}$$

where  $A$  is a diagonal matrix. We must show that  $\ddagger_N$ , the normalized covariance matrix of  $\ddagger$ , is identical to  $\ddagger_y$ . Normalizing  $\ddagger$  we have

$$\begin{aligned}\ddagger_N &= D\ddagger D \\ &= D (A \ddagger_x A) D\end{aligned}$$

where  $D$  is again diagonal but in this case,

$$d_{ii} = \frac{1}{\sqrt{\sigma_{ii} a_{ii}^2}}$$

with  $a_{ii}$  the  $ii^{\text{th}}$  element of diagonal matrix  $A$ . For the  $ij^{\text{th}}$  element of  $\ddagger_N$  we have

$$\begin{aligned}\sigma_{Nij} &= d_{ii} a_{ii} \sigma_{ij} a_{jj} d_{jj} \\ &= \frac{a_{ii} \sigma_{ij} a_{jj}}{\sqrt{\sigma_{ii} a_{ii}^2} \sqrt{\sigma_{jj} a_{jj}^2}} \\ &= \sigma_{ij},\end{aligned}$$

Thus, this procedure of normalization makes the entries of the normalized covariance matrix invariant with respect to translating and scaling transformations on the grey tone  $N$ -tuples.

**APPENDIX IV**  
**COMPUTER PROGRAM DOCUMENTATION & LISTINGS**

**IV.1 PROGRAM DOCUMENTATION**

- IV.1-a ERTS Retrieval Programs**
- IV.1-b Texture Analysis Programs**
- IV.1-c Cross-Band Texture Analysis Programs**
- IV.1-d Piecewise Linear Classification Programs**

**IV.2 PROGRAM LISTINGS**

- IV.2-a ERTS Retrieval Programs**
- IV.2-b Texture Analysis Programs**
- IV.2-c Cross-Band Texture Analysis Programs**
- IV.2-d Piecewise Linear Classification Programs**



IV.1-a ERTS Retrieval Programs - Documentation

RETV

ERTS

PIXEY

ZEQUAN

PITCHR

WRTDSK

RDDSK1

KEQUAN

RDDSK2

## ERTS RETRIEVAL PROGRAMS

REPRODUCIBILITY OF THE  
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PROGRAM TITLE: RETV  
VERSION: 1  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

This is the mainline for the ERTS Retrieval Programs which retrieve ERTS MSS data from standard NASA digital tapes and outputs the image data in a picture, or greytone listing, or copies it onto an output tape.

### INPUT PARAMETERS under NAMELIST 'PARAM':

NBAND	Band number, 1 through 4 for MSS band 4 through 7, to be selected. Set to 5 for all 4 bands, assumed to be 2.
IRSTRT, ICSTRT	Row, column starting coordinates.
IRSTOP, ICSTOP	Row, column stopping coordinates.
TITLE	80 Column title for output list.
MILLI	TRUE if coordinates are in millimeters; assumed to be FALSE, coordinates are row, column points in ERTS image.
SMALL	TRUE if millimeter coordinates are from a small 70 mm by 70 mm negative; assumed to be FALSE, coordinates from a 7 inch by 7 inch print.
PRNT	TRUE for grey tone listing; assumed FALSE.
TAPE	TRUE for tape output; assumed FALSE.

IFIL	Output file code for tape; assumed to be 02.
PIC	TRUE for picture output; assumed TRUE.
QUAN	TRUE for equals probability quantizing of the image; assumed to be TRUE.
SPIC	TRUE for special picture run-see PIXEY; assumed to be FALSE.

#### REQUIREMENTS:

1. ERTS input tape must be on file code 'ES'.
2. Four disc files must be on files 11, 12, 13 and 14.
3. Any output tape must be positioned on file code 'IFIL'.
4. All coordinates must be determined relative to the input tape rather than the print -- that is, ICSTOP must not exceed 824 points or 46 mm.
5. Core - 22 k
6. Subprograms required:

RETV  
 ERTS  
 PIXEY  
 ZEQUAN  
 ERTS  
 PITCHR  
 WRTDSK  
 RDDSKI  
 KEQUAN  
 PITCHR  
 RDDSK2

#### COMMENTS:

For efficiency, data is read by RETV in blocks of 41 lines by 41 points. One ERTS tape (one-fourth of an image) will then be covered by 20 horizontal blocks, leaving 4 points left at the end of each line. Note --- sometimes on the first tape of an image, the first four points are greytone of 255 and can adversely affect a picture printout. If so, set ICSTRT = 5.

Also note that if the point ICSTRT is not a multiple of 41 from the end of the line, then the last points may not necessarily be listed since blocks are determined starting from ICSTRT.

A special picture run can be made to print out and reduce any image over the entire tape by setting SPIC to TRUE and specifying under namelist PARAM the following parameters: IRSTRT, IRSTOP, ICSTRT, ICSTOP, NBAND, QUAN. Then under namelist PICTUR, parameters for PITCHR are specified. See PIXEY for details.

## ERTS RETRIEVAL PROGRAM

SUBPROGRAM TITLE: ERTS  
VERSION: II  
DATE: September, 1972  
UPDATE: November, 1973  
AUTHOR: G. Gunnels  
DOCUMENTED BY: R. Bosley  
PROGRAM LANGUAGE: GMAP  
IMPLEMENTED ON: HW635  
PURPOSE:

To read 7-track ERTS MSS data tapes.

### ENTRY POINTS:

CALL EINIT (NOLS)  
CALL ESKIP (NOSK)  
CALL EREAD (I, LN)  
CALL EREWIND

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### ARGUMENTS:

NOLS Number of words per scan line; returned by EINIT.  
NOSK The number of records to skip.  
I The array into which the NOLS words of data  
from a line of ERTS data is placed.  
LN Returned by EREAD giving the line number of the  
line of data returned. If LN = 0, the end of file  
was reached on the ERTS tape.

### ERROR FLAGS:

MB EREAD buffer is not large enough for a block of  
ERTS data.  
AI EINIT was called twice.  
NI EINIT was not called before calling EREAD,  
ESKIP, or EREWIND.

UE EOF encountered while reading ID or  
annotation blocks on ERTS tape.  
EF EOF encountered while trying to skip records  
in ESKIP.

**COMMENTS:**

EINIT initializes the ERTS tape so that data may be read, and must be called first. ESKIP skips over NOSK records (scan lines). EREWIND rewinds the ERTS tape. EINIT must not be called twice. The data placed into array I by EREAD is in standard corresponding point forms. Since the ERTS MSS data has four channels, there are actually  $NOLS/4$  points or cells per scan line.

**REQUIREMENTS:**

ERTS tape must be on file code 'ES'.

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## ERTS RETRIEVAL PROGRAMS

SUBPROGRAM TITLE: PIXEY  
VERSION: 1  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

To provide a flexible option for printing a picture of ERTS MSS data using user specified parameters to PITCHR.

### ENTRY POINT:

CALL PIXEY (ILINE, IMAGE, IRSTRT, IRSTOP, ICSTRT, ICSTOP, QUAN, NBAND)

### INPUT ARGUMENTS:

ILINE	Array the ERTS line is read into.
IMAGE	Array containing one line of the image.
IRSTRT, IRSTOP	Starting, stopping row in the image.
ICSTRT, ICSTOP	Starting, stopping column in the image.
QUAN	TRUE for equal probability quantizing of the image.
NBAND	Band number to be processed; set to 5 for all 4 MSS bands.

### INPUT PARAMETERS: under NAMELIST 'PICTUR':

LNSKIP,	Line and column increment for ERTS data;
KOLSKP	assumed to be 4,3.

**ICELL, JCELL** Number of rows, columns in image to be printed;  
 assumed to be 1,256.  
**INIT** Number of times entry to PITCHR is made at  
 SNAP: must be greater than 1; assumed to be 304.  
**IMIN, IMAX** Minimum, maximum greytone in image;  
 assumed to be 0,75 or 12 if QUAN is true .  
**NROW** Number of rows to be printed; equal to ICELL.  
**NFILES** Number of output files available to PITCHR;  
 set to 0 for all output on file code 06, set to 2  
 for files 06 and 42; assumed to be 2.  
**IFIL(10)** Array containing output file codes; assumed  
 to be 06 and 42.  
**NULW, NULD** Number of columns, rows per output page;  
 assumed to be 129, 60.  
**AMAG, DMAG** Width, length magnification for output picture;  
 assumed to be 1.

**SAMPLE RUN:**

\$PARAM SPIC=T, QUAN=T, IRSTRT=1, IRSTOP=1216,  
 ICSTRT=5, ICSTOP=772, NBAND=2\$END/\$PICTUR INIT=304,  
 NFILES=2, LNSKIP=4, KOLSKP=3, JCELL=256\$END.

This run will print out on files 06 and 42 a picture 256 points  
 wide by 304 lines long. Note that  $1216/4 = 304$  and  $786/3 = 256$   
 gives the values for LNSKIP=4, KOLSKP=3, JCELL=256,  
 INIT=304. Also, using these values for LNSKIP and KOLSKP  
 will result in a picture in proportion to the same area on an ERTS  
 image print, approximately twelve 64 by 64 subimages across  
 by 19 subimages down the tape.

**COMMENTS:**

If SPIC=T in the \$PARAM card, then a \$PICTUR card must follow.  
 Note that the parameters IRSTRT, IRSTOP, ICSTRT, ICSTOP,  
 QUAN, NBAND are supplied on the \$PARAM card while LNSKIP,  
 KOLSKP and all PITCHR parameters are supplied on the \$PICTUR  
 card.

The program will output a negative picture but a positive picture can be made by setting  $IMAX=0$  and  $IMIN=75$  (or 12 if quantization is used). The parameters are initialized to output an area approximately twelve 64 by 64 subimages wide by 19 down. For a complete description of PITCHR parameters, see PITCHR.

REQUIREMENTS:

1. Processor time for one band with parameters as shown in the sample run is 0.075 hr with 3k lines of output.
2. Subprograms required are ZEQUAN, PITCHR.

CALLED BY:

RETV

## ERTS RETRIEVAL PROGRAM

SUBPROGRAM TITLE: ZEQUAN  
VERSION: I  
DATE: September, 1973  
UPDATE: September, 1973  
AUTHOR: Z. Dinstein  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

To equal probability quantize a large image on disc to NQ levels.

### ENTRY POINT:

CALL ZEQUAN (LINE, NUMLIN, NUMPPL, NCOMP, ICOMP,  
LEFT, NQ, INFIL, IOUTFL)

### INPUT ARGUMENTS:

LINE	Array to store one line of the image.
NUMLIN	Number of lines in the image.
NUMPPL	Number of columns in the image.
NCOMP	Number of components in the image.
ICOMP	The component to be quantized.
LEFT	Left-most cell desired in the line.
NQ	Number of quantized levels.
INFIL	File code of disc containing the image to be quantized.

### OUTPUT ARGUMENTS:

IOUTFL	Output file code for the quantized image.
--------	---

COMMENTS:

Processing is done line by line after an initial pass through the image is made to determine the number of grey tone levels in the image. The minimum and maximum grey tones are printed. The number of grey tone levels should not exceed 512. Input data on disc INFILE must be in binary.

CALLED BY:

PIXEY

## ERTS RETRIEVAL PROGRAM

SUBPROGRAM TITLE: PITCHR  
VERSION: II  
DATE: July, 1969  
UPDATE: November, 1970  
AUTHOR: R. Cowles  
DOCUMENTED BY: G. Elliott  
PROGRAM LANGUAGE: GMAP  
IMPLEMENTED ON: HW635  
PURPOSE:

To print out images in 13 grey levels.

### ENTRY POINTS:

CALL PITCHR (IRRAY, ICELL, JCELL, INIT, IT, IMIN, IMAX,  
NROW, NFIL, IFIL, NULW, NULD, AMAG, DMAG, \*)  
CALL SNAP  
CALL SNAPA (IARRAY)  
CALL ENDBNR

### ARGUMENTS:

IRRAY	Array to be printed, either integer or floating point.
ICELL	Number of rows in array. (row dimension)
JCELL	Number of columns in array. (column dimension)
INIT	=0 if all of image to be printed out is in core at time of call. Output will be done before return to calling program. =1 for reinitialization entry. Any of the arguments previously specified with INIT = 0 with the exception of INIT may now be changed. Return will be made to the calling program without any output. This is especially useful if the image is read into core in pieces and the last piece does not completely fill the array. >1 for initialization

entry. INIT will reflect the number of times entry is made at SNAP before final border is to be printed. Return is to the calling routine without any output.  
 INIT =0 is assumed.

II =0 if array is floating point  
 =1 if array is integer  
 II =1 is assumed.

IMIN Minimum brightness level in array. Type of IMIN should correspond to that indicated by II.  
 IMIN =0 is assumed.

IMAX Maximum brightness level in array. Type of IMAX should correspond to that indicated by II.  
 IMAX =12 is assumed.

NROW Number of rows of array to be printed if full array is not to be printed. This allows for partial printing.  
 NROW = ICELL is assumed.

NFIL Number of output files available if image is to be output in strips that are NULW lines wide.  
 =0 for all output on file code 06  
 =1 for all output on file code IFIL  
 1 for outputs in strips, on file codes specified in array IFIL. NFIL=0 is assumed.

IFIL Array containing output file codes. Ignored unless NFIL = 0. IFIL must be a variable rather than a literal, since SNAP alters the value(s) of IFIL to contain the location of the relevant file control block in the upper half of the word.

NULW Number columns per output page. MAX =129  
 NULW =120 unless otherwise specified.

NULD Number of rows to be printed before a slew on the top of the next page is given. NULD = ICELL \* DMAG + 1.



AMAG

Floating point magnification in width.

AMAG=1 is assumed.

DMAG

Floating point magnification in length,

DMAG=1 is assumed.

\*

Error return if not enough output files are available. Return is made with NARG=-10. If not specified, a message will be printed out on the accounting report and an NF report will terminate execution.

## ERTS RETRIEVAL PROGRAMS

SUBPROGRAM TITLE: WRTDSK  
VERSION: II  
DATE: September, 1972  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

To write one line of ERTS data onto disc files, one file for each band.

### ENTRY POINT:

CALL WRTDSK (ILINE, NHOR, IPSTR, IPEND, NBAND)

### INPUT ARGUMENTS:

ILINE	Array containing one line of ERTS data
NHOR	Number of horizontal blocks of 41 columns in the line.
IPSTR, IPEND	Starting and ending points in the ERTS line.
NBAND	The desired band; set to 5 for all 4 MSS bands.

### COMMENTS:

After reading a line of data into ILINE, RETV calls WRTDSK to pick out the segment of NHOR blocks in the line and write it onto disc. If only one band is desired, only that band is put onto disc. Disc files 11, 12, 13, and 14 must be present.

### CALLED BY:

RETV

## ERTS RETRIEVAL PROGRAMS

SUBPROGRAM TITLE: RDDSKL  
VERSION: II  
DATE: September, 1972  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

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To read the ERTS data from disc and print out a picture of the image, proceeding vertically and then from left to right.

### ENTRY POINT:

CALL RDDSKL (IMAGE, QUAN, NHOR, NVERT, NDSK)

### INPUT ARGUMENTS:

IMAGE	Array used to store two 41 x 41 blocks.
QUAN	TRUE for equal probability quantization of the image.
NHOR	Number of horizontal 41 x 41 blocks.
NVERT	Number of vertical 41 x 41 blocks.
NDSK	File code of disc to be processed.

### ERROR FLAGS:

Same as for PITCHR.

### COMMENTS:

Since blocks are 41 columns wide, then two horizontal blocks are processed together, except for the final strip of blocks when NHOR is an odd number. It is recommended that QUAN be set to true since transmission errors result in very high and very low greytone along the line which will make the true image features indistinguishable unless equal probability quantizing is used.

SUBPROGRAMS CALLED:

KEQUAN

PITCHR

CALLED BY:

RETV

## ERTS RETRIEVAL PROGRAMS

SUBPROGRAM TITLE: KEQUAN  
VERSION: II  
DATE: September, 1971  
UPDATE: June, 1973  
AUTHOR: G. Elliot  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

To equal probability quantize the input array to NQ levels.

### ENTRY POINT:

CALL KEQUAN (IA, NGL, NQ, IASIZE)

### ARGUMENTS:

IA	Input array which is returned quantized.
NGL	Number of greytone levels in IA.
NQ	Number of quantizing levels.
IASIZE	Size of array IA.

### ERROR FLAGS:

If the number of greytone levels exceeds 512, an error message is printed.

### CALLED BY:

RDDSK1

## ERTS RETRIEVAL PROGRAMS

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SUBPROGRAM TITLE: RDDSK2  
VERSION: II  
DATE: September, 1972  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

To read ERTS data from disc and print out the grey tones and/or copy the data onto an output tape.

### ENTRY POINT:

CALL RDDSK2 (IMAGE, IRSTRT, IRSTOP, NHOR, NDSK,  
PRNT, TAPE, IFIL)

### INPUT ARGUMENTS:

IMAGE	Array data is read into.
IRSTRT, IRSTOP	Starting, stopping lines of data.
NHOR	Number of horizontal blocks of 41 columns.
NDSK	File code of the disc to be read.
PRNT	TRUE for greytone listing.
TAPE	TRUE for tape output
IFIL	Output tape file code.

### COMMENTS:

If neither the grey tone listing nor the output tape is desired, execution is returned to the calling program, RETV.

### CALLED BY:

RETV

IV.1-b Texture Analysis Programs - Documentation

MAINLN

ERTS

MAING

KEQUAN

PITCHR

FPLXIT

INDEX

IMOMTR

COR

IEQPQ1

RITOWT



## ERTS TEXTURE ANALYSIS

C-2

PROGRAM TITLE: MAINLN  
VERSION: II  
DATE: September, 1972  
UPDATE: June, 1973  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Craig Paul  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To read the Texture Analysis parameters, read in subimages from ERTS input tape, and send these subimages to MAING for processing.

### INPUT PARAMETERS:

1. Title Card: Up to 80 columns for use as identification information.
2. Parameters under NAMELIST /PARAM/:
  - IBAND Band used for processing-
    - = 1 for MSS Band 4
    - = 2 for MSS Band 5
    - = 3 for MSS Band 6
    - = 4 for MSS Band 7assumed to be 2.
  - NUMPPL Number of points per line of each subimage; assumed to be 64.
  - NUMLIN Number of lines of each subimage; assumed to be 64.
  - NBVERT Number of the last vertical subimage to be processed; assumed to be 36.
  - NUMIM Number of subimages taken horizontally in one run; assumed to be 3. Note:  
NUMPPL x NUMIM must not exceed 192 points.

C-2

**N11** The upper-left column coordinate for the vertical strip (of NUMIM horizontal subimages) being processed in this run.

**NUMSTR** The number of the vertical strip being processed; assumed to be 1.

**NBSKIP** The number of vertical rows of subimages to skip before beginning processing; assumed to be 0.

**PNCH** Specifies the output option:  
 = 1HY for card output  
 = 1HT for tape output on file code IF  
 = 1HN for neither card nor tape but a listing of the LEX arrays; assumed to be tape (1HT).  
 Note: PNCH must be denoted as a Hollerith constant in the \$PARAM card.

**IF** Output tape file code which is assumed to be in position (it is not rewound); assumed to be '03'.

**NRED** The base used for image reduction; assumed to be 1.

**NSTART, NTIMES** Each subimage is processed NSTART through NTIMES times, each time with a new reduction factor  $NFT = NRED ** N LAYER$  where N LAYER goes from NSTART - 1 through NTIMES - 1; both assumed to be 1.

**PICTUR** TRUE for a picture of each subimage; assumed FALSE.

**NQUANT** Number of quantization levels of the probability function in IMOMTR; assumed to be 16.

**MERGE** TRUE for all four LEX arrays to be merged into one array; assumed to be TRUE.

REQUIREMENTS:

1. ERTS input tape must be on file code 'ES'.
2. Core - 43k.
3. Disc file must be on file code '02'.
4. Subprograms required:

MAINLN

ERTS

MAING

KEQUAN

PITCHR

FPLXIT

INDEX

IMOMTR

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COR

IEQPQ1

RITOWT

CARD SETUP FOR SAMPLE RUN:

```
$ IDENT      9999, ANYNAME
$ LIBRARY    LB
$ OBJECT     ERTS TEXTURE ANALYSIS PROGRAMS
$ DKEND
$ EXECUTE
$ PRMFL      LB, R, S, PATTERN/GEE/LIB
$ TAPE       ES, A5DD, 60500, ERTS00, INPUT
$ FILE       02, A2R, 2L
$ LIMITS     20, 43k, 10k
$ INCODE     IBMF
```

TEST-SETUP FOR TEXTURE ANALYSIS PROGRAMS

```
$ PARAM      N11=1, PNCH = 1HN, PICTUR = TSEND
$ ENDJOB
```

This run of the texture analysis programs will process the ERTS image in 64 x 64 subimages, giving only printed output plus a picture of each subimage.

## COMMENTS:

This is the mainline of the texture analysis programs. Each ERTS image is divided into 4 vertical strips, each put onto a 7-track digital tape. This image is divided up into subimages for processing. For example, if the subimages are 64 columns by 64 lines, then the first tape contains subimages 1 through 12, the second contains 13 through 24, and so on up to 48 horizontal subimages. And, each tape contains 36 vertical subimages. Often, the first 8 points at the beginning of each line have grey tones of 255 and will adversely affect processing. Therefore, the first eight points of each line are skipped.

Each tape contains usually 3296 points of 4 bands, or 824 points per line for one band. Skipping the first eight points leaves 816 points per line. If subimages are each 64 columns, then 12 will occupy 768 columns, leaving 48 points at edge of each tape unused.

Due to core limitations, each input tape is processed in vertical strips of up to 192 horizontal points. This is 3,64 by 64 subimages, or 6,32 by 32 subimages in one run. For example, if NUMPPL = 64 and NUMIM = 3, then the tape is processed as follows: Run 1 - (1, 1), (1, 2), (1, 3), (2, 1), (2, 2), (2, 3), (3, 1), ..., (36, 1), (36, 2), (36, 3). Run 2 - (1, 4), (1, 5), (1, 6), (2, 4), (2, 5), (2, 6), (3, 4), ..., (36, 4), (36, 5), (36, 6). Run 3 - (1, 7), (1, 8), (1, 9), (2, 7), (2, 8), (2, 9), (3, 7), ..., (36, 7), (36, 8), (36, 9). Run 4 - (1, 10), (1, 11), (1, 12), ..., (36, 10), (36, 11), (36, 12). This completes the processing of the first tape. Note that N11 is determined relative to the entire image while NUMSTR is relative to the input tape. That is, for tape 1, N11 and NUMSTR are both 1 for Run 1. But for Run 2, N11 is 4 while NUMSTR is 2 and for Run 3, N11 is 7 while NUMSTR is 3, etc.

Continuing with the example, tape 2 will be processed as follows: Run 1 (N11= 13, NUMSTR=1) - (1, 13), (1, 14), (1, 15), ..., (36, 13), (36, 14), (36, 15). Run 2 (N11=16, NUMSTR=2) - (1, 16), (1, 17), (1, 18), ..., (36, 16), (36, 17), (36, 18), etc.

See Figure 1 for an illustration of an ERTS image divided into 64 by 64 subimages. Note - the size of each subimage must not exceed 4096 points.

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### ERTS IMAGE

### 4 Tapes for Each ERTS Image

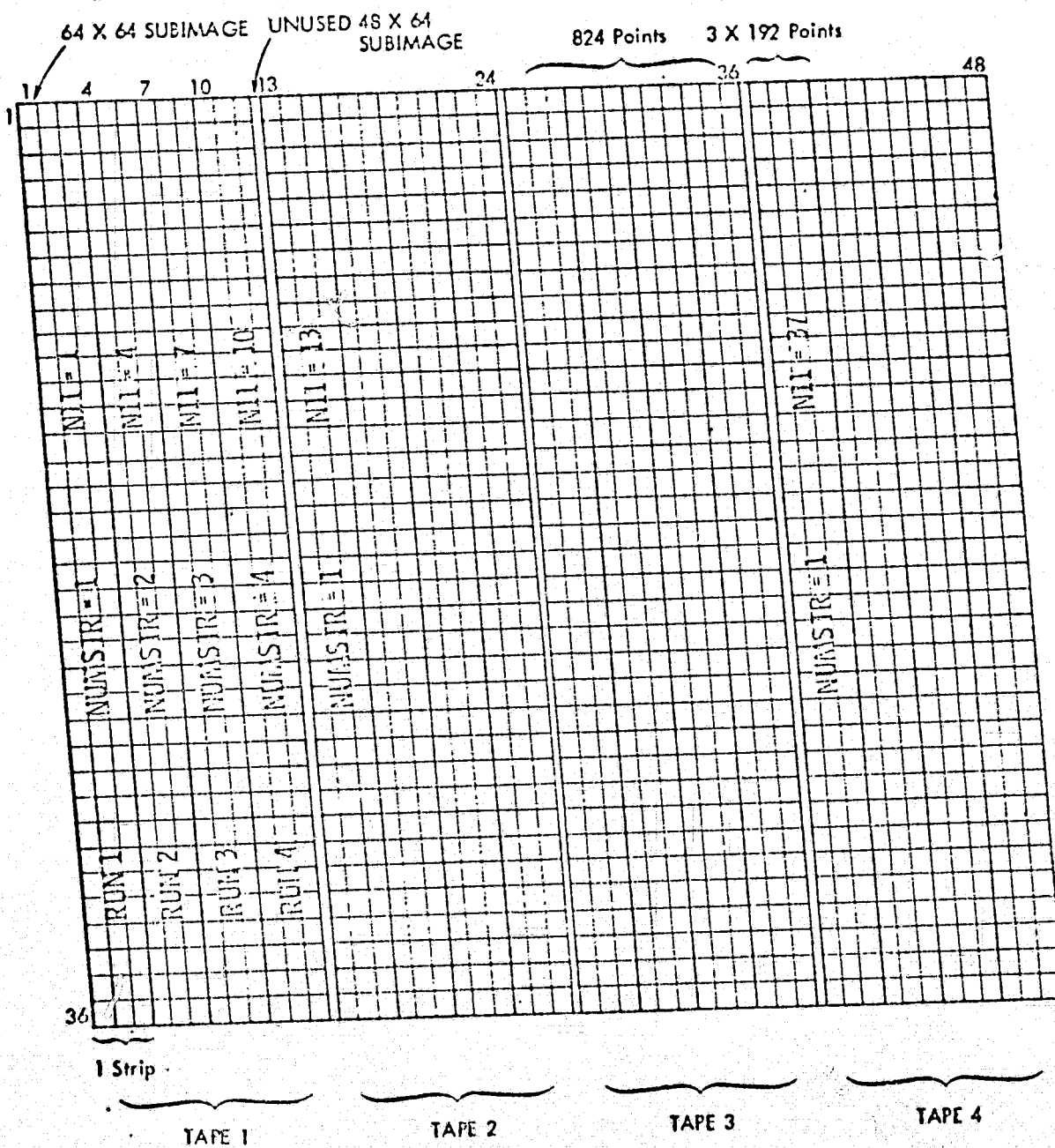


Figure 1.

SUBPROGRAM TITLE:	ERTS
VERSION:	II
DATE:	September, 1972
UPDATE:	November, 1973
AUTHOR:	G. Gunnels
DOCUMENTED BY:	R. Bosley
PROGRAM LANGUAGE:	GMAP
IMPLEMENTED ON:	HW635
PURPOSE:	

To read 7-track ERTS MSS data tapes.

#### ENTRY POINTS:

CALL EINIT (NOLS)  
 CALL ESKIP (NOSK)  
 CALL EREAD (I, LN)  
 CALL EREWIND

#### ARGUMENTS:

NOLS	Number of words per scan line; returned by EINIT.
NOSK	The number of records to skip.
I	The array into which the NOLS words of data from a line of ERTS data is placed.
LN	Returned by EREAD giving the line number of the line of data returned. If LN=0, the end of file was reached on the ERTS tape.

#### ERROR FLAGS:

MB	EREAD buffer is not large enough for a block of ERTS data.
AI	EINIT was called twice.
NI	EINIT was not called before calling EREAD, ESKIP, or EREWIND.



UE EOF encountered while reading ID or  
annotation blocks on ERTS tape.  
EF EOF encountered while trying to skip records  
in ESKIP.

**COMMENTS:**

EINIT initializes the ERTS tape so that data may be read, and must be called first. ESKIP skips over NOSK records (scan lines). EREWND rewinds the ERTS tape. EINIT must not be called twice. The data placed into array I by EREAD is in standard corresponding point forms. Since the ERTS MSS data has four channels, there are actually NOLS/4 points or cells per scan line.

**REQUIREMENTS:**

ERTS tape must be on file code 'ES'.

## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: MAING  
VERSION: III  
DATE: October, 1971  
UPDATE: June, 1973  
AUTHOR: R. M. Haralick  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To print out a picture of the subimage, copy it to file 2, and then process it through the texture subroutines FPLXIT to get the LEX arrays, IMOMTR to calculate the texture features, and RITOWT to output the results.

### ENTRY POINT:

CALL MAING (IWORK, MERR, MERGE, PICTUR, IF)

### ARGUMENTS:

IWORK	NUMLIN by NUMPPL subimage array.
MERR	Set to 1 if an error occurs in fitting the LEX arrays into IWORK.
MERGE	TRUE to merge the four LEX arrays into one.
PICTUR	TRUE for a picture of the subimage.
IF	File code for the output tape.

### COMMENTS:

The subimage sent in IWORK is scaled to fill a page for the picture printout, and at the same time it is copied to a scratch disc on file code '02'. If the size of the LEX arrays is greater than NUMPPL x NUMLIN, then an error message is printed and processing is terminated by MAINLN, after putting EOF marks on the output tape. Each subimage is quantized by KEQUAN to 32 levels.

CALLED BY:

MAINLN

SUBPROGRAMS REQUIRED:

KEQUAN

PITCHR

FPLXIT

IMOMTR

RITOWT

SUBPROGRAM TITLE: KEQUAN  
VERSION: II  
DATE: September, 1971  
UPDATE: June, 1973  
AUTHOR: G. Elliot  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW635  
PURPOSE:

To equal probability quantize the input array to NQ levels.

ENTRY POINT:

CALL KEQUAN (IA, NGL, NQ, IASIZE)

ARGUMENTS:

IA	Input array which is returned quantized.
NGL	Number of greytone levels in IA.
NQ	Number of quantizing levels.
IASIZE	Size of array IA.

ERROR FLAGS:

If the number of greytone levels exceeds 512, an error message is printed.

SUBPROGRAM TITLE: PITCHR  
 VERSION: II  
 DATE: July, 1969  
 UPDATE: November, 1970  
 AUTHOR: R. Cowles  
 DOCUMENTED BY: G. Elliott  
 PROGRAM LANGUAGE: GMAP  
 IMPLEMENTED ON: HW635  
 PURPOSE:

To print out images in 13 grey levels.

ENTRY POINTS:

CALL PITCHR (IRRAY, ICELL, JCELL, INIT, IT, IMIN, IMAX,  
 NROW, NFIL, IFIL, NULW, NULD, AMAG, DMAG,\*)  
 CALL SNAP  
 CALL SNAPA (IARRAY)  
 CALL ENDBNR

ARGUMENTS:

IRRAY Array to be printed, either integer or floating point.  
 ICELL Number of rows in array. (row dimension)  
 JCELL Number of columns in array. (column dimension)  
 INIT =0 if all of image to be printed out is in core at time of call. Output will be done before return to calling program.  
 =1 for reinitialization entry. Any of the arguments previously specified with INIT =0 with the exception of INIT may now be changed. Return will be made to the calling program without any output. This is especially useful if the image is read into core in pieces and the last piece does not completely fill the array. >1 for initialization

entry. INIT will reflect the number of times entry is made at SNAP before final border is to be printed. Return is to the calling routine without any output.  
 INIT =0 is assumed.

II =0 if array is floating point  
 =1 if array is integer  
 II =1 is assumed.

IMIN Minimum brightness level in array. Type of IMIN should correspond to that indicated by II.  
 IMIN =0 is assumed.

IMAX Maximum brightness level in array. Type of IMAX should correspond to that indicated by II.  
 IMAX =12 is assumed.

NROW Number of rows of array to be printed if full array is not to be printed. This allows for partial printing.  
 NROW = ICELL is assumed.

NFIL Number of output files available if image is to be output in strips that are NULW lines wide.  
 =0 for all output on file code 06  
 =1 for all output on file code IFIL  
 >1 for outputs in strips, on file codes specified in array IFIL. NFIL=0 is assumed.

IFIL Array containing output file codes. Ignored unless NFIL = 0. IFIL must be a variable rather than a literal, since SNAP alters the value(s) of IFIL to contain the location of the relevant file control block in the upper half of the word.

NULW Number columns per output page. MAX =129  
 NULW =120 unless otherwise specified.

NULD Number of rows to be printed before a slew on the top of the next page is given. NULD = ICELL \* DMAG + 1.

AMAG Floating point magnification in width.  
AMAG =1 is assumed.

DMAG Floating point magnification in length,  
DMAG =1 is assumed.

\* Error return if not enough output files are  
available. Return is made with NARG =-10.  
If not specified, a message will be printed out  
on the accounting report and an NF report  
will terminate execution.



## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: FPLXIT  
VERSION: II  
DATE: September, 1971  
UPDATE: June, 1973  
AUTHOR: R. M. Haralick  
DOCUMENTED BY: R. J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To compute the four nearest neighbor greytone matrices LEX1, LEX2, LEX3, LEX4, for angles of 90, 0, 135, and 45 degrees.

### ENTRY POINT:

CALL FPLXIT (IDATA, LEX1, LEX2, LEX3, LEX4, NUMPPL, MERGE)

### ARGUMENTS:

IDATA	Scratch array holding two lines of the subimage.
LEX1 - LEX 4	Address indices for the four LEX arrays.
NUMPPL	Number of points in each line of the subimage.
MERGE	TRUE to merge the four LEX arrays into one array.

### COMMENTS:

This subroutine reads two lines at a time from the subimage being processed, which is now on scratch disc file 02. After all the LEX arrays are created, they can be merged into one by setting MERGE to TRUE in MAINLN. The merge is performed by adding each array term by term and putting the total into LEX1.

SUBPROGRAMS REQUIRED:

INDEX

CALLED BY:

MAING

## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: INDEX  
VERSION: I  
DATE: September, 1971  
UPDATE: September, 1971  
AUTHOR: R. M. Haralick  
DOCUMENTED BY: R. J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To return to the calling program the single subscript for the LEX array that indicates where element (I, L) can be found, given its row and column subscripts I and L.

### ENTRY POINT:

INDEX (I, J)

### ARGUMENTS:

I

Row subscript for an element in the LEX array.

J

Column subscript for an element in the LEX array.

### COMMENTS:

This subprogram is a FUNCTION.

### CALLED BY:

FPLXIT

IMOMTR

## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: IMOMTR  
VERSION: II  
DATE: September, 1971  
UPDATE: June, 1973  
AUTHOR: R. M. Haralick  
DOCUMENTED BY: R. J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To calculate the moment texture statistics.

### ENTRY POINT:

CALL IMOMTR (LEX1, LEX2, LEX3, LEX4, F, IQ, MERGE)

### ARGUMENTS:

LEX1 - LEX4 Address indices for the four LEX arrays.  
F Cumulative distribution function.  
IQ Quantized output array of IEQPQ1,  
MERGE TRUE indicates the four LEX arrays have been merged into one.

### TEXTURE FEATURES:

#### 1. Angular Second Moment

$$\text{ANGMOM} = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} P^2(i, j)$$

where  $N_g$  is the number of grey tone levels, and  $P(i, j)$  is the array of joint probabilities.

2. Entropy

$$\text{ENTROP} = - \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} P(i, j) \log P(i, j)$$

3. Mean

$$\text{AMEAN} = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} i \cdot P(i, j)$$

4. Variance

$$\text{SGMASQ} = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} (i - \text{AMEAN})^2 P(i, j)$$

5. Covariance

$$\text{SGMAXY} = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} (i - \text{AMEAN}) (j - \text{AMEAN}) P(i, j)$$

6. Correlation

$$\text{RATIO} = \frac{\text{SGMAXY}}{\text{SGMASQ}}$$

7. Inverse Moment

$$\text{IVDMOM} = \sum_{i=1}^{N_g} \sum_{j=1}^{N_g} \frac{P(i, j)}{1 + (i-j)^2}$$

8. Average Contrast

$$\text{DIFAVE} = \sum_{k=1}^{N_g-1} k \cdot \text{DIF}(k)$$

$$\text{where DIF}(k) = \sum_{|i-j|=k} P(i, j)$$

9. Variance of DIF

$$\text{DIFVAR} = \sum_{k=1}^{N_g-1} k^2 \cdot \text{DIF}(k) - \left[ \sum_{k=1}^{N_g-1} k \cdot \text{DIF}(k) \right]^2$$

10. Entropy of DIF

$$\text{DIFENT} = - \sum_{k=1}^{N_g-1} \text{DIF}(k) \cdot \log(\text{DIF}(k))$$

11. Average of Intensity

$$\text{SUMAVE} = \sum_{k=2}^{2N_g} k \cdot \text{SUM}(k)$$

$$\text{where SUM} = \sum_{i+j=k} P(i, j)$$

12. Variance of SUM

$$\text{SUMVAR} = \sum_{k=2}^{2N_g} k^2 \cdot \text{SUM}(k) - \left[ \sum_{k=2}^{2N_g} k \cdot \text{SUM}(k) \right]^2$$

13. Entropy of SUM

$$\text{SUMENT} = - \sum_{k=2}^{2N_g} \text{SUM}(k) \cdot \log(\text{SUM}(k))$$

14. True mean of probability function

$$\text{TMEAN} = \frac{1}{N_g} \sum_{i=1}^{N_g} F(i)$$

COMMENTS:

The three remaining texture features are computed in subroutine COR: CORINF, CORMUT, and CORMAX. If MERGE is TRUE, then these features are computed for only the merged array, LEX1. Otherwise they are computed for each LEX array, corresponding to each of four angles.

SUBPROGRAMS REQUIRED:

INDEX  
IEQPQ1  
COR

CALLED BY:

MAING



## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: COR  
VERSION: II  
DATE: November, 1972  
UPDATE: June, 1973  
AUTHOR: Sam Shanmugam  
DOCUMENTED BY: R. J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To calculate three measures of correlation between two discrete random variables X and Y whose joint probabilities of occurrence are known.

### ENTRY POINT:

CALL COR(PXY, N, IOPT, Q, COR1, COR2, COR3)

### INPUT ARGUMENTS:

PXY      Array of joint probabilities.  
N        Size of the array PXY  
IOPT     Option flag - if IOPT = 0, then only COR1 and COR2 will be calculated; if IOPT = 1, then COR3 will also be calculated.  
Q        Scratch array of size N x N. This array is used only if IOPT is non-zero.

### OUTPUT ARGUMENTS:

COR1     Maximal correlation measure.  
COR2     Information measure of correlation.  
COR3     Second type of maximal measure.

### COMMENTS:

These three correlation measures are the last three texture features. For details of the measures see "Mutual Information and Maximal Correlation As Measure of Dependence," by C. B. Bell, in the Annals of Mathematical Statistics, vol. 43, 1962.

### CALCULATIONS:

$$1. \text{COR1} = \frac{H(x, y) - H_1(x, y)}{\max(H(x), H(y))}$$

$$\text{where } H(x, y) = \sum_i \sum_j \log(p_{xy}^2(i, j))$$

$$H_1(x, y) = \sum_i \sum_j (\log(p_x(i) p_y(j))) p_{xy}(i, j)$$

$$H(x) = \sum_i (\log p_x(i)) p_x(i)$$

$$\text{and } H(y) = \sum_j (\log p_y(j)) p_y(j)$$

$$2. \text{COR2} = \sqrt{1 - e^{-2R}}$$

$$\text{where } R = H_2(x, y) - H(x, y)$$

$$H_2(x, y) = \sum_i \sum_j (\log p_x(i) p_y(j)) p_x(i) p_y(j)$$

$$\text{and } p_x(i) = \sum_j p_{xy}(i, j), p_y(j) = \sum_i p_{xy}(i, j).$$

3. COR3 is computed using the eigenvector corresponding to the second largest eigenvalue of  $QQ^T$ , where

$$Q(I, J) = \frac{P_{xy}(i, j)}{\sqrt{P_x(i) P_y(j)}}$$

CALLED BY:

IMOMTR

## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: IEQPQ1  
VERSION: 1  
DATE: September, 1971  
UPDATE: September, 1971  
AUTHOR: D. Goel  
DOCUMENTED BY: R. J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

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To determine k levels of equal probability quantization for an array for which the cumulative distribution function is known for all elements.

### ENTRY POINT:

CALL IEQPQ1 (N, K, F, IQ, IMIN)

### ARGUMENTS:

N	Number of items in array F to be equal probability quantized.
K	Number of quantizing levels.
F	Input array to be quantized.
IQ	Output array of quantized levels.
IMIN	The lowest possible level in the input data.

### CALLED BY:

IMOMTR

## ERTS TEXTURE ANALYSIS

SUBPROGRAM TITLE: RITOWT  
VERSION: II  
DATE: September, 1971  
UPDATE: June, 1973  
AUTHOR: R. M. Haralick  
DOCUMENTED BY: R. J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To output onto printer, cards, or tape the texture features.

### ENTRY POINT:

CALL RITOWT (LEX1, LEX2, LEX3, LEX4, G, IQ, MERGE,  
IF, PICTUR)

### ARGUMENTS:

LEX1 -LEX4	Address indices for the LEX arrays.
G	CDF for the image data
IQ	Quantized output of IEQPQ1 of NQUANT levels.
MERGE	TRUE indicates that the four LEX arrays have been merged into one array.
IF	File code for output tape.
PICTUR	TRUE indicates that a picture of the subimage has been printed.

### COMMENTS:

The output format for the listing is slightly different depending upon the PICTUR and merge options. The PNCH option determines whether cards or tape or neither are used to output the 17

texture features for each subimage. In any case, the texture features are listed on the printer. If PNCH = 1HY in the \$PARAM card in MAINLN, then cards are punch according to the following formats:

1. for MERGE = TRUE: M1, N1, NFT, ANGMOM, ENTROP, RATIO, SGMASQ, SGMAXY, AMEAN, VIDMOM, KOUNT/TMEAN, DIFENT, DIFAVE, DIFVAR, SUMENT, SUMAVE, SUMVAR, KOUNT+1/CORINF, CORMUT, CORMAX, KOUNT+2.

FORMAT (I2, 1X, 2I2, 1X, 7F9.5, 19/8X, 7F9.5, 19/8X, 3F9.5, 36X, 19), where (M1, N1) is the subimage row, column coordinate.

2. for MERGE = FALSE: M1, N1, NFT/ANGMOM(4), ENTROP(4), KOUNT/RATIO(4), SGMASQ(4), KOUNT+1/SGMAXY(4), AMEAN(4), KOUNT+2/VIDMOM(4), TMEAN(4), KOUNT+3/DIFENT(4), DIFAVE(4), KOUNT+4/DIFVAR(4), SUMENT(4), KOUNT+5/SUMAVE(4), SUMVAR(4), KOUNT+6/CORINF(4), CORMUT(4), KOUNT+7/CORMAX(4), KOUNT+8, where (4) denotes four values, one for each angle.

FORMAT ('THE SCENE(', I2, 1, ', ', I2, ') HAS BEEN REDUCED BY', 15/8(1X, 8F9.5, 17/)/1X, 4F9.4, 38X, 15).

If tape output on file code 'IF' is selected by PNCH = 1HT, then the texture features are written in binary as follows:

WRITE(IF) M1, N1, NFT, ANGMOM(K), ENTROP(K), RATIO(K), SGMASQ(K), SGMAXY(K), AMEAN(K), VIDMOM(K), TMEAN(K), DIFENT(K), DIFAVE(K), DIFVAR(K), SUMENT(K), SUMAVE(K), SUMVAR(K), CORINF(K), CORMUT(K), CORMAX(K), where K is one for MERGE = TRUE and is 4 for MERGE = FALSE denoting the number of values for each measure.

If neither cards nor tape output is selected, then the LEX arrays are listed after the texture features.

CALLLED BY:

MAING

IV.1-c Cross-Band Texture Analysis Programs -  
Documentation

SPECTR

GETIM / GETIT

ERTS

DIFFER

COVAR

MNCVIN / MNCV

CORREL

## CROSS-BAND TEXTURE ANALYSIS

PROGRAM TITLE: SPECTR  
VERSION: I  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTY  
IMPLEMENTED ON: HW635  
PURPOSE:

This program is the mainline of the spectral-textural analysis which obtain spectral-textural features for land-use classification of ERTS MSS data.

### ENTRY POINT:

CALL SPECTR (IMAGE, X, ILINE, IXDIM, IYDIM, NDIN)

### INPUT ARGUMENTS:

IMAGE            Array containing a subimage.  
                  (IYDIM, IXDIM,  
                  IDIN)  
X (IYDIM,        Array containing the difference image.  
                  IXDIM, IDIN)  
ILINE            Array one ERTS line is read into.  
IXDIM            Column dimension of IMAGE, X.  
IYDIM            Row dimension of IMAGE, X.  
NDIN             Number of components, bands, in IMAGE, X.

### INPUT PARAMETERS: under NAMELIST 'PARAM':

NDIM            Number of components desired in IMAGE:  
                  Assumed to be NDIN .  
NUMLIN         Number of lines in subimage; Assumed to be IYDIM.



NUMPPL	Number of columns in subimage; Assumed to be IYDIM.
FMT	Format used to output elements of covariance matrix; assumed to be 'E11.4'.
TITLE	80 column title for run.
OPT	TRUE to print covariance matrix; assumed FALSE.
IDIST	Distance between neighboring cells for difference array; assumed to be 1.
IRSTRT	Starting row in ERTS image; assumed to be 1.
IRSTOP	Stopping row in ERTS image; assumed as last row.
LAPHOR	Number of horizontal points that subimages overlap; assumed to be 0.
LAPVER	Number of vertical points that subimages overlap; assumed to be 0.
PNCH	TRUE for output on cards, FALSE for output to file code 01 for tape or disc.

EXAMPLE OF DRIVER:

```

DIMENSION IMAGE (16, 17, 8), X (16, 17, 8), ILINE (3300)
EQUIVALENCE (IMAGE, X, ILINE(130))
IXDIM =17
IYDIM =16
NDIN =8
CALL SPECTR (IMAGE, X, ILINE, IXDIM, IYDIM, NDIN)
STOP
END

```

This driver will set up the spectral-textural analysis mainline SPECTR to process 16 x 16 subimages over 8 components with IDIST=1. Note IXDIM must include NUMPPL plus IDIST points, and array ILINE must have at least NUMPPL x NDIM points outside of any other array. These points form array XLINE in COVAR.

#### REQUIREMENTS:

1. Core - 25k for IMAGE (32, 33, 8)
2. ERTS input tape must be on file code 'ES'
3. Random access disc file on file 11, eg. \$ FILE II AIR, OR.
4. Subprograms required:
  - DRIVER
  - SPECTR
  - GETIM
  - SETDIM (Fortran callable program to initialize HEMP package)
  - GETIT
  - ERTS
  - DIFFER
  - COVAR
  - MNCVIN
  - MNCV
  - CORREL
  - SFA07F
  - HEMDET (Fortran callable program from HEMP library to solve for determinant of matrix).

#### COMMENTS:

These analysis programs obtain a series of NUMLIM by NUMPPL by NDIM subimages from the ERTS input tape and outputs a feature vector with  $(1 + NDIM (NDIM-1)/2)$  components for each subimage. The input data is processed in horizontal rows of subimages that may overlap both horizontally and vertically. The distance between neighboring resolution cells used to get the difference array is variable. Note that IXDIM must include NUMPPL plus IDIST points, and that the array ILINE must have at least NUMPPL \* NDIM points outside of any other array because these are used for array XLINE in COVAR. Other than this, arrays IMAGE, X, and ILINE may be equivalenced

to conserve core, as in the example for a DRIVER. The first feature component on the output file is the entropy measure, and the remaining  $NDIM (NDIM-1)/2$  components are elements of the correlation matrix. See the GETIM subprogram for a listing of all 8 possible components for a subimage.

CALLED BY:

DRIVER

## CROSS-BAND TEXTURE ANALYSIS

SUBPROGRAM TITLE: GETIM  
VERSION: I  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTY  
IMPLEMENTED ON: HW635  
PURPOSE:

To get a row of subimages from the ERTS input tape and copy them onto a random access disc file.

### ENTRY POINTS:

CALL GETIM (ILINE, IDIST, NDIM, IRSTRT, IRSTOP, NUMLIN,  
NUMPPL, LAPVER, LAPHOR, NHOR, INCR, IPEND)  
CALL GETIT

### INPUT ARGUMENTS:

ILINE	Array into which one ERTS line is read.
IDIST	Distance between neighboring resolution cells whose differences form the difference array, X.
NDIM	Number of components in each resolution cell.
IRSTRT	Starting line of ERTS data.
IRSTOP	Stopping line of ERTS data.
NUMLIN	Number of lines in each subimage.
NUMPPL	Number of points per line (columns) in each subimage.
LAPVER	Number of lines that subimages overlap.
LAPHOR	Number of columns that subimages overlap.

OUTPUT ARGUMENTS:

NHOR            Number of horizontal overlapping subimages in image.  
INCR            Horizontal increment to the first column of the  
                  next subimage in the row.  
IPEND            Last cell in the row.

REQUIREMENTS:

Random access disc file on file code II: and FILE II, AIR, IOR.

COMMENTS:

This program is initialized by calling GETIM which initializes the ERTS input tape and sets up the disc on file II for random access with fixed length records of NDIM words, up to a maximum of  $NDIM = 8$ . The eight possible greytone components are:

1. Band 2
2. Band 3
3. Band 4
4. Band 2 x Band 2
5. Band 3 x Band 3
6. Band 4 x Band 4
7. Band 2 x Band 3
8. Band 2 x Band 4, where band 1 through 4 is MSS band 4 through 7.

It is suggested that all eight components be used and the best of these be selected for feature vector components. This gives a total of 29 components, including the entropy measure.

After calling GETIM, all further calls are made to GETIT which goes down the input data file line by line copying to the random access disc an entire row of NHOR subimages of NDIM components. Note that MSS band 4 (band 1 here) has been deleted because of its high correlation with MSS band 5 (band 2 here).

CALLED BY:

SPECTR

SUBPROGRAM TITLE: ERTS  
VERSION: II  
DATE: September, 1972  
UPDATE: November, 1973  
AUTHOR: G. Funnels  
DOCUMENTED BY: R. Bosley  
PROGRAM LANGUAGE: GMAP  
IMPLEMENTED ON: HW635  
PURPOSE:

To read 7-track ERTS MSS data tapes.

ENTRY POINTS:

CALL EINIT (NOLS)  
CALL ESKIP (NOSK)  
CALL EREAD (I, LN)  
CALL EREWND

ARGUMENTS:

NOLS	Number of words per scan line; returned by EINIT.
NOSK	The number of records to skip.
I	The array into which the NOLS words of data from a line of ERTS data is placed.
LN	Returned by EREAD giving the line number of the line of data returned. If LN=0, the end of file was reached on the ERTS tape.

ERROR FLAGS:

MB	EREAD buffer is not large enough for a block of ERTS data.
AI	EINIT was called twice.
NI	EINIT was not called before calling EREAD, ESKIP, or EREWND.

UE EOF encountered while reading ID or  
annotation blocks on ERTS tape.  
EF EOF encountered while trying to skip records  
in ESKIP.

**COMMENTS:**

EINIT initializes the ERTS tape so that data may be read, and must be called first. ESKIP skips over NOSK records (scan lines). EREWND rewinds the ERTS tape. EINIT must not be called twice. The data placed into array I by EREAD is in standard corresponding point forms. Since the ERTS MSS data has four channels, there are actually NDLS/4 points or cells per scan line.

**REQUIREMENTS:**

ERTS tape must be on file code 'ES'.

## CROSS-BAND TEXTURE ANALYSIS

SUBPROGRAM TITLE: DIFFER  
VERSION: 1  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTY  
IMPLEMENTED ON: HW635  
PURPOSE:

To calculate the nearest neighbor difference array.

### ENTRY POINT:

CALL DIFFER (IA, X, IXDIM, IYDIM, NDIM, IDIST, NUMPPL,  
NUMLIN)

### INPUT ARGUMENTS:

IA(IYDIM, IXDIM, NDIM) Array containing the subimage being processed.  
IXDIM Column dimension of IA and X.  
IYDIM Row dimension of IA and X.  
NDIM Number of components of each resolution cell.  
NUMPPL Number of columns in the subimage.  
NUMLIN Number of lines in the subimage.

### OUTPUT ARGUMENTS:

X(IYDIM, IXDIM, NDIM) Array of nearest neighbor differences.



COMMENTS:

This subroutine will replace the original subimage in array IA with the nearest neighbor horizontal difference:  $(I_1 - J_1, I_2 - J_2, \dots, I_N - J_N)$  where I and J are N-dimensional horizontally neighboring resolution cells separated by distance IDIST. Arrays IA and X may be equvalenced to occupy the same area of core. Note that the absolute value is used to get the differences. This gives only the positive half of the distribution of differences I-J and J-I. This shifts the mean of the distribution from the origin and must be accounted for in COVAR when the covariance matrix of the difference array is calculated.

CALLED BY:

SPECTR

## CROSS-BAND TEXTURE ANALYSIS

SUBPROGRAM TITLE: COVAR  
VERSION: 1  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTY  
IMPLEMENTED ON: HW635  
PURPOSE:

To calculate the covariance matrix of the difference array.

### ENTRY POINT:

CALL COVAR (XLINE, NDIM, NUMPPL, X, IXDIM, IYDIM,  
NUMLIN, NDIN, COV)

### INPUT ARGUMENTS:

XLINE(NDIM, NUMPPL)	Array used to send one line of the difference array to MNCVIN.
NDIM	Number of components of each vector in X.
NUMPPL	Number of columns of vectors in X.
X	Nearest neighbor difference array.
IXDIM, IYDIM	Column, row dimensions of array X.
NUMLIN	Number of rows of vectors in X.
NDIN	Dimension of COV array.

### OUTPUT ARGUMENTS:

COV (NDIN, NDIN)	Covariance matrix of the difference array X.
---------------------	--

### REQUIREMENTS:

Subroutine MNCVIN.

COMMENTS:

Array XLINE is formed from the first NDIM x NUMPPL points of array ILINE in SPECTR. Hence at least the first NDIM x NUMPPL words of ILINE must not be equvalenced into array X. Since only the positive differences were used to make array X by DIFFER, the mean is reset to zero for each component .

CALLED BY:

SPECTR

## CROSS-BAND TEXTURE ANALYSIS

SUBPROGRAM TITLE: MNCVIN  
VERSION: 1  
DATE: August, 1973  
UPDATE: August, 1973  
AUTHOR: James D. Young  
DOCUMENTED BY: James D. Young  
PROGRAM LANGUAGE: FORTRAN IV or FORTY  
IMPLEMENTED ON: HW635  
PURPOSE:

To calculate the mean vector and covariance matrix for each category of a set of vectors, based on a specified percentage of the vectors randomly chosen within the set.

### ENTRY POINTS:"

CALL MNCVIN (NVPCAL, NDIM, NCALL, PERCNT, NCAT,  
X, NTRUTH, COV, XMEAN, SCTMEN, SAMSZ, IERROR, JERROR)

CALL MNCV

### INPUT ARGUMENTS:

NVPCAL	Number of vectors per call.
NDIM	Dimension of data vectors.
NCALL	Number of calls.
PERCNT	Percentage of total number of vectors from which the mean and covariance matrices will be calculated.
NCAT	Number of categories considered; set to 1 if only one set of statistics will be calculated for all data, set to the number of categories in data set if one set of statistics will be calculated for each category.

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X (NDIM, NVPICAL) Matrix containing input data vectors in its columns.

NTRUTH (NVPICAL) Vector containing the ground truth integers, 1 through NCAT, associated with the data vectors of X; if NCAT is 1 this vector is ignored.

OUTPUT ARGUMENTS:

COV (NDIM, NDIM, NCAT) Matrix containing covariance matrices of the data.

XMEAN (NDIM, NCAT) Matrix containing mean vectors of the data.

SCTMEN (NDIM, NCAT) Scratch matrix.

SAMSZ (NCAT) Vector with the number of vectors used to calculate the statistics for each category.

IERROR Error flag when returned non-zero:

1. if NVPICAL .LE. 0
2. if NDIM .LE. 0
3. if NCAL .LE. 0
4. if PERCENT .GT. 100. or so small that less than 2 vectors will be used to calculate all the statistics.
5. if NCAT .LE. 0

JERROR Error flag when returned non-zero:

1. if an illegal ground truth label is formed.

COMMENTS:

One call to MNCVIN initializes this routine. Calls to MNCV should be performed NCALL times, each with the next line of vectors in X. After MNCV has been called NCALL times, the mean vector and covariance matrix for each category is completed. For use in the Spectral-Textural Analysis programs, PERCENT is set to 100 and NCAT is set to 1.

## CROSS-BAND TEXTURE ANALYSIS

SUBPROGRAM TITLE: CORREL  
VERSION: I  
DATE: January, 1974  
UPDATE: January, 1974  
AUTHOR: Robert J. Bosley  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTY  
IMPLEMENTED ON: HW 635  
PURPOSE:

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To calculate the correlation matrix given the covariance matrix  
of the difference array.

### ENTRY POINT:

CALL CORREL (COV, NDIM, COR)

### INPUT ARGUMENTS:

COV           Covariance matrix of the difference array  
NDIM           Order of matrix COV.

### OUTPUT ARGUMENTS:

COR           Correlation matrix of COV.

### CALLED BY:

SPECTR

IV.1-d Piecewise Linear Classification Programs -  
Documentation

RCLASS  
XIN  
LINEAR  
WEIGHT

## LINEAR DISCRIMINANT CLASSIFIER

SUBPROGRAM TITLE: RCLASS  
VERSION: I  
DATE: September, 1972  
UPDATE: November, 1972  
AUTHOR: Sam Shanmugam  
DOCUMENTED BY: Sam Shanmugam  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

This subroutine is the mainline for routines which implement a decision rule using piecewise linear discriminant functions. It calls subroutine XIN to read in the pattern vectors, and calls LINEAR to get the decision rule and classify the pattern vectors.

### ENTRY POINT:

CALL RCLASS (WORK, ISIZE)

### ARGUMENTS:

WORK Scratch array of size ISIZE which holds the training vectors and the weight vectors.  
ISIZE Dimension of array WORK.

### INPUT PARAMETERS:

Card 1. Format (4I1)

NOPT1 Set to 1 to print out training patterns; otherwise set to zero.  
NOPT2 Set to 1 to print out test patterns; otherwise set to zero.  
NOPT3 Set to 0 to list only the contingency table for the training patterns; otherwise set to 1 and the classification of each training pattern is listed as well as the contingency table.



NOPT4 Set to 0 for only the contingency table of the test set; otherwise set to 1 for the classification of each test pattern as well.

Card 2. Format (515)

NTOT The total number of pattern vectors in the data set.

NPART NPART out of every ten pattern vectors in the data set will be used for training. The remaining will be used as test patterns.

NDIM The number of measurements per vector plus two.

NC The number of ground truth categories.

NPAIR Twice the maximum number of training patterns in any one category.

#### REQUIREMENTS:

1. Maximum number of categories is 15.
2. Maximum number of components, NDIM, is 100.
3. Pattern vectors must be sorted by category.
4. A scratch disc file must be on 02.
5. ISIZE must be at least

$$\begin{aligned} & \text{NDIM (NTRAIN + 10) + 1000} \\ & + (\text{NC (NC + 1)/2}) \text{ND} + \text{NPAIR} \cdot \text{ND} \\ & \text{where ND = NDIM - 1} \end{aligned}$$

and NTRAIN = number of training patterns.

6. Pattern vectors must be written in binary to disc file 01 as follows:

WRITE (01) IGT, MI, NI, NFT, (FEAT(I), I=1, NMEAS)

where

IGT is the ground truth category.

MI, NI, NFT are not used - may be used as ID tags

FEAT is the feature vector

NMEAS is the number of measurements per feature vector.

## 7. Subprograms required

DRIVER

RCLASS

XIN

LINEAR

WEIGHT

### ERROR FLAGS:

If ISIZE is too small, processing is terminated and an error message is listed.

### THEORY:

Using a regression type algorithm the program obtains a set of hyperplanes for separating the training patterns of different category pairs. A total of  $NC(NC-1)/2$  hyperplanes are determined. Test patterns are identified by taking a majority vote on this set of hyperplanes. For complete details, see "Introduction to Statistical Pattern Recognition" by Y. Fukunaga, Academic Press, 1972.

### COMMENTS:

The input data file on 01 should have a total of NTOT logical records in binary. Each logical record must be of length  $NDIM + 2$  words, where word 1 is the ground truth category and words set is sorted into training and test sets according to NPART. Training vectors are stored in WORK and test vectors are copied to disc 02.

CALL BY:

DRIVER

EXAMPLE OF DRIVER:

```
DIMENSION WORK (10000)
ISIZE = 10000
CALL RCLASS (WORK, ISIZE)
STOP
END
```

## LINEAR DISCRIMINANT CLASSIFIER

SUBPROGRAM TITLE: XIN  
VERSION: II  
DATE: September, 1972  
UPDATE: December, 1973  
AUTHOR: Sam Shanmugam  
DOCUMENTED BY: Robert J. Bosley  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To read the parameter cards and the input data set, copying the test vectors to disc.

ENTRY POINT:

CALL XIN (WORK, U)

ARGUMENTS:

WORK      Array training vectors are read into  
U          Scratch array

CALLED BY:

RCLASS

COMMENTS:

This version does not use subroutine POST to position the input file 01. Hence the input data file must be the first file on 01.

## LINEAR DISCRIMINANT CLASSIFIER

SUBPROGRAM TITLE: LINEAR  
VERSION: 1  
DATE: September, 1972  
UPDATE: November, 1972  
AUTHOR: Sam Shanmugam  
DOCUMENTED BY: Sam Shanmugam  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

Using the set of training patterns, this program obtains a set of hyperplanes for pairwise separation of training patterns of different categories. The program also identifies the test patterns on a majority vote on the hyperplanes and outputs contingency table.

### ENTRY POINT:

CALL LINEAR (XTRAIN, XTEST, W, U, DUMMY)

### ARGUMENTS:

XTRAIN	Matrix containing training patterns.
XTEST	Matrix containing test patterns.
W	Array of weight vectors.
U	Matrix used for calculating the boundary between category pairs.
DUMMY	Scratch array.

### CALLED BY:

RCLASS

SUBPROGRAMS REQUIRED:  
WEIGHT

## LINEAR DISCRIMINANT CLASSIFIER

SUBPROGRAM TITLE: WEIGHT  
VERSION: I  
DATE: September, 1972  
UPDATE: November, 1972  
AUTHOR: Sam Shanmugam  
DOCUMENTED BY: Sam Shanmugam  
PROGRAM LANGUAGE: FORTRAN IV  
IMPLEMENTED ON: HW 635  
PURPOSE:

To find the minimum mean square fit hyperplane for separating the training patterns of two different categories.

### ENTRY POINT:

CALL WEIGHT (U, DUMMY, WT, ND, NIJ)

### INPUT ARGUMENTS:

U Array containing the patterns of category I and category J.  
DUMMY Scratch array.  
ND ND+1 is the dimension of the pattern vector.  
NIJ Number of vectors in U.

### OUTPUT ARGUMENTS:

WT Weight vector which defines the hyperplane separating categories I and J.

### CALLED BY:

LINEAR

### SUBPROGRAMS REQUIRED:

MINV

Matrix inversion program from the IBM Scientific Subroutine Package.

IV.2-a ERTS Retrieval Program Listings

RETV

ERTS

PIXEY

ZEQUAN

PITCHR

WRTDSK

RDDSK1

KEQUAN

RDDSK2

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02-12-74

18.974

ERTS RETRIEVAL PROGRAM

CRETV

ERTS RETRIEVAL PROGRAM

VERSION 1

WRITTEN BY RJ BOSLEY

JAN 1974

DESCRIPTION OF PROGRAM.

THIS IS THE MAINLINE FOR ERTS RETRIEVAL PROGRAMS WHICH WILL RETRIEVE DATA FROM STANDARD NASA ERTS DATA TAPES AND OUTPUT THE IMAGE DATA IN A PICTURE, OR GREY-TONE LISTING, OR COPY IT ONTO AN OUTPUT TAPE ON FILE CODE 'IFIL'. COORDINATES MAY BE EITHER IN MILLIMETERS OR IN ROW, COLUMN COUNT. FOR REASONABLE EFFICIENCY DATA IS READ IN BLOCKS OF 41 LINES BY 41 POINTS. ONE ERTS TAPE WILL THEN BE COVERED BY 21 HORIZONTAL BLOCKS, LEAVING 4 POINTS LEFT AT THE END OF EACH LINE. NOTE--SOMETIMES ON THE FIRST TAPE OF AN IMAGE, THE FIRST FOUR POINTS ARE 255 AND WILL ADVERSELY AFFECT A PICTURE PRINTOUT. IF SO, SKIP THE FIRST FOUR POINTS BY SETTING ICSTRT=5.

NOTE---IF THE POINT ICSTRT IS NOT A MULTIPLE OF 41 FROM THE END OF THE LINE, THEN THE LAST POINTS MAY NOT NECESSARILY BE OUTPUTTED SINCE THE BLOCKS ARE DETERMINED BY THE ICSTRT POSITION. THEREFORE IT IS SUGGESTED THAT IF THE DATA OUT TO THE END OF EACH LINE IS DESIRED, THAT THE COORDINATES BE SPECIFIED IN ROW,COLUMN FORMAT AND THAT THEY BE PREDETERMINED SO AS TO OUTPUT THE DESIRED POINTS IN THE ERTS LINE.

A SPECIAL PICTURE RUN TO PRINT OUT AND REDUCE ANY IMAGE OVER THE ENTIRE TAPE CAN BE MADE BY SETTING SPIC TO .TRUE. IN \$PARAM AND THEN SPECIFYING PICTURE PARAMETERS UNDER NAMELIST \$PICTUR. SEE THE PIXEY PROGRAM FOR AN EXAMPLE OF A SAMPLE RUN.

INPUT PARAMETERS UNDER NAMELIST /PARAM/.

NBAND

BAND NUMBER TO BE SELECTED

SET=5 FOR ALL BANDS

IRSTRT,ICSTRT

ROW,COLUMN STARTING COORDINATES

IRSTOP,ICSTOP

ROW,COLUMN STOPPING COORDINATES

NHOR

NUMBER OF HORIZONTAL 41 X 41 BLOCKS THAT COVERS AREA SELECTED

NVERT

NUMBER OF VERTICAL BLOCKS COVERING AREA SELECTED

IPSTR,IPEND

STARTING,ENDING POINTS IN ERTS LINE

TITLE

TITLE FOR THIS RUN

IFIL

OUTPUT FILE FOR TAPE, SET TO 02

ILINE

ARRAY WHERE ERTS LINE IS READ INTO

IMAGE

ARRAY TO STORE 41 X 41 SUBIMAGE BLOCK

PICT

TRUE FOR PICTURE OUTPUT

SPIC

TRUE FOR SPECIAL PICTURE RUN

QUAN

TRUE FOR QUANTIZED PICTURE OUTPUT

PPNT

TRUE FOR GREY-TONE LISTING

TAPE

TRUE FOR TAPE OUTPUT

MILLI

TRUE IF COORD ARE SPECIFIED IN MM.

SMALL

TRUE IF MM. COORD COME FROM A SMALL

70 X 70 NEGATIVE---OTHERW. SE ASSUMED

RETV0001  
RETV0002  
RETV0003  
RETV0004  
RETV0005  
RETV0006  
RETV0007  
RETV0008  
RETV0009  
RETV0010  
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RETV0041  
RETV0042  
RETV0043  
RETV0044  
RETV0045  
RETV0046  
RETV0047  
RETV0048  
RETV0049  
RETV0050



TO BE FALSE--FROM A 7 X 7 INCH PRINT

RESTRICTIONS.

1. ERTS INPUT TAPE MUST BE ON FILE CODE 'ES'.
2. FOUR JISC FILES MUST BE ON FILES 11,12,13,14.
3. ANY OUTPUT TAPE MUST BE POSITIONED ON FILE CODE 'IFIL'.
4. ALL COORDINATES MUST BE DETERMINED RELATIVE TO THE INPUT TAPE---THAT IS, ICSTOP MUST NOT EXCEED 824 COLUMNS OR 46MM. STARTING COORDINATES MUST BE AT LEAST 1.

SUBPROGRAMS REQUIRED.

```

RETV
  ERTS (WITH EREWND)
  PIXEY
    ZEQUAN
    ERTS
    PITCHR
  WRTOSK
  RDSK1
    KEQUAN
    PITCHR
  RDSK2
    
```

```

DIMENSION IMAGE(41,82), ILINE(3300), TITLE(14)
EQUIVALENCE (IMAGE(1,1), ILINE(1))
LOGICAL EOF, PIC, SPIC, QUAN, PRNT, TAPE, SMALL, MILLI
DATA BLANK/' ' /
NAMelist /PARAM/NBAND, PTC, QUAN, PRNT, TAPE, IRSTRT, ICSTRT, IRSTOP,
1 ICSTOP, IFIL, TITLE, MILLI, SPIC
    
```

\*\*\*\*\* SECTION I --- SET UP PARAMETERS FOR PROCESSING \*\*\*\*\*

INITIALIZE ERTS AND EOF FLAG

```

CALL EINIT(LENGTH)
CALL FLGEOF(05, EOF)
    
```

INITIALIZE PARAMETERS

```

1001 PIC=.TRUE.
    SPIC=.FALSE.
    QUAN=.TRUE.
    PRNT=.FALSE.
    TAPE=.FALSE.
    SMALL=.FALSE.
    MILLI=.FALSE.
    DO 7 I=1,14
7   TITLE(I)=BLANK
    IFIL=2
    NBAND=2
    IRSTRT=1
    
```

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RETVO051  
RETVO052  
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RETVO100

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02-12-74 18.974 ERTS RETRIEVAL PROGRAM

	ICSTRT=1		RETV0101
	IPSTOP=0		RETV0102
	ICSTOP=0		RETV0103
C		READ IN PARAMETERS	RETV0104
	READ(05,PARAM)		RETV0105
C		CHECK FOR END OF CARDS	RETV0106
	IF(EOF) STOP		RETV0107
C		WRITE OUT THE PARAMETERS	RETV0108
	WRITE(6,1) (TITLE(I),I=1,14),NBAND		RETV0109
	IF(NBAND.EQ.5) WRITE(6,2)		RETV0110
	IF(TAPE) WRITE(6,3) IFIL		RETV0111
	IF(PIC) WRITE(6,4) QUAN		RETV0112
	WRITE(6,5) IRSTRT,IRSTOP,ICSTRT,ICSTOP		RETV0113
1	FORMAT(20X,'ERTS RETRIEVAL PROGRAM      VERSION II'///1X,14A6////		RETV0114
1	' BAND NUMBER IS ',I3)		RETV0115
2	FORMAT(' ALL FOUR ERTS BANJS WILL BE PROCESSED')		RETV0116
3	FORMAT(' AN OUTPUT TAPE WILL BE CREATED ON FILE ',I3)		RETV0117
4	FORMAT(' PICTURE QUANTIZATION IS ',L1)		RETV0118
5	FORMAT(' STARTING ROW IS    ',I6,10X,'ENDING ROW IS    ',I6/		RETV0119
1	' STARTING COLUMN IS   ',I6,10X,'ENDING COLUMN IS   ',I6)		RETV0120
C		CHANGE MM COORD INTO ROW AND COL	RETV0121
	IF(.NOT.M.LLI) GO TO 19		RETV0122
	WRITE(6,6)		RETV0123
6	FORMAT(' COORDINATES ARE IN MILLIMETERS')		RETV0124
C		GET THE CORRECT IMAGE SIZE	RETV0125
	IF(SMALL) GO TO 10		RETV0126
C		ASSUME THAT A PRINT HAS 180MM VERT	RETV0127
C		AND 184 MM ACROSS OR 2336 ROWS VERT	RETV0128
C		AND 3296 POINTS ACROSS.	RETV0129
C		HOWEVER--SINCE THE TAPE ONLY COVERS	RETV0130
C		ONE FOURTH ACROSS AN IMAGE, THEN USE	RETV0131
C		46MM ACROSS AND 824 COLUMNS ACROSS.	RETV0132
	IRSTRT=IPSTRT*2336/180-12		RETV0133
	IRSTOP=IRSTOP*2336/180		RETV0134
	ICSTRT=ICSTRT*824/46-17		RETV0135
	ICSTOP=ICSTOP*824/46		RETV0136
	GO TO 15		RETV0137
C		DO SMALL NEGATIVE COORDINATES	RETV0138
		A NEGATIVE IS 70MM BY 70MM	RETV0139
C			RETV0140
10	IRSTRT=IRSTRT*2336/70-31		RETV0141
	IRSTOP=IRSTOP*2336/70		RETV0142
	ICSTRT=ICSTRT*824*2/35-46		RETV0143
	ICSTOP=ICSTOP*824*2/35		RETV0144
C		WRITE OUT THE NEW COORDINATES	RETV0145
15	WRITE(6,16) IRSTRT,IRSTOP,ICSTRT,ICSTOP		RETV0146
16	FORMAT('           COORDINATES IN ROW AND COLUMN FORMAT ARE---'/		RETV0147
1	'           STARTING ROW IS    ',I6,10X,'ENDING ROW IS    ',I6/		RETV0148
1	'           STARTING COLUMN IS ',I6,10X,'ENDING COLUMN IS ',I6)		RETV0149
C		CHECK COORDINATES SO THEY FALL IN	RETV0150
C		ONE TAPE	

-12-74

18.974

ERTS RETRIEVAL PROGRAM

```

19 IF(ICSTOP.GT.824) GO TO 20 CHECK PARAMETERS RETV0151
C IF(NBAND.GT.5) GO TO 20 RETV0152
IF(NBAND.LT.1) GO TO 20 RETV0153
IF(IRSTRT.LT.1) GO TO 20 RETV0154
IF(ICSTRT.LT.1) GO TO 20 RETV0155
IF(ICSTOP.LT.IRSTRT) GO TO 20 RETV0156
IF(ICSTOP.GT.ICSTRT) GO TO 22 RETV0157
20 WRITE(6,21) RETV0158
21 FORMAT(///' ***EXECUTION TERMINATED FOR THIS RUN--ERROR IN PARAMET RETV0159
1ERS'/1+1) RETV0160
GO TO 1001 RETV0161
C CHECK FOR SPECIAL PICTURE RJN RETV0162
22 IF(.NOT.SPIC) GO TO 18 RETV0163
CALL PIXEY(ILINE,LINE,IRSTRT,IRSTOP,ICSTRT,ICSTOP,QUAN,NBAND) RETV0164
GO TO 1001 RETV0165
C POSITION THE INPUT TAPE RETV0166
18 NOSK=IRSTPT-1 RETV0167
CALL EREWND RETV0168
IF(NOSK.NE.0) CALL ESKIP(NOSK) RETV0169
WRITE(6,23) LENGTH RETV0170
23 FORMAT(' LENGTH OF ONE ERTS LINE IS ',I6) RETV0171
IF(LENGTH.LE.3300) GO TO 25 RETV0172
WRITE(6,24) RETV0173
24 FORMAT(///' LENGTH OF LINE ON ERTS EXCEEDS DIMENSION OF ARRAY ILINE RETV0174
1E---EXECUTION TERMINATED') RETV0175
STOP RETV0176
C DETERMINE THE NUMBER OF BLOCKS RETV0177
C ALLOW 15 PTS TO BE CUT OFF BEFORE RETV0178
C STARTING A NEW STRIP OF BLOCKS RETV0179
25 NHOR=((ICSTOP-ICSTRT-15)/41)+1 RETV0180
C RESET ENDING COLUMN RETV0181
27 ICSTOP=NHOR*41+ICSTRT-1 RETV0182
IF(ICSTOP.LE.824) GO TO 26 RETV0183
NHOR=NHOR-1 RETV0184
GO TO 27 RETV0185
26 NVERT=((IRSTOP-IRSTRT-15)/41)+1 RETV0186
IPSTOP=41*NVERT+IRSTRT-1 RETV0187
C SET STARTING AND ENDING PTS IN ILINE RETV0188
IPSTR=(ICSTRT-1)*4 RETV0189
IPEND=ICSTOP*4 RETV0190
C WRITE OUT NO OF BLOCKS AND STOPS RETV0191
28 WRITE(6,28) NHOR,NVERT,IRSTOP,ICSTOP RETV0192
FORMAT(' NUMBER OF HORIZONTAL BLOCKS IS ',I5/ RETV0193
1 ' NUMBER OF VERTICAL BLOCKS IS ',I5/ RETV0194
1 ' REVISED STOPPING ROW IS NOW ',I6/ RETV0195
1 ' REVISED STOPPING COLUMN IS NOW ',I6) RETV0196
C ***** SECTION II --- PROCESS THE ERTS DATA TAPE ***** RETV0197
C RETV0198
C RETV0199
C RETV0200

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C			RETV0201
C		PART 1--READ THE DATA	RETV0202
C		GO THRU EACH VERTICAL BLOCK	RETV0203
	DO 29 ID=11,14		RETV0204
29	REWIND ID		RETV0205
	DO 100 LINE=IRSTRT,IRSTOP		RETV0206
	CALL EPEAD(ILINE,LN)		RETV0207
C		CHECK FOR ERROR	RETV0208
	IF(LN.NE.0) GO TO 90		RETV0209
	WRITE(6,80)LN		RETV0210
80	FORMAT(////' ***EXECUTION TERMINATED---EOF DETECTED ON ERTS TAPE')		RETV0211
	GO TO 301		RETV0212
C		WRITE NHOR RECORDS ON DISC, ONE DISC	RETV0213
C		PER BAND	RETV0214
C		CALL WRTDSK(ILINE, NHOR, IPSTR, PEND, NBAND)	RETV0215
90	CONTINUE		RETV0216
100		PART 2--OUTPUT THE DATA	RETV0217
C		SET UP NO OF TIMES AND DISC F.LE CODE	RETV0218
C			RETV0219
	NTIMES=1		RETV0220
	NDSK=10+NBAND-1		RETV0221
	IF(NBAND.NE.5) GO TO 250		RETV0222
	NTIMES=4		RETV0223
	NDSK=10		RETV0224
C		GO THRU ONE TIME FOR EACH BAND	RETV0225
250	DO 300 ITIME=1,NTIMES		RETV0226
	NDSK=NDSK+1		RETV0227
C		OUTPUT THE PICTURE FOR THIS BAND	RETV0228
	IF(PIC) CALL RDSK1(IMAGE, QUAN, NHOR, NVERT, NDSK)		RETV0229
		WRITE OUT THE GREY-TONES	RETV0230
C		CALL RDSK2(IMAGE,IRSTRT,IRSTOP,NHOR,NDSK,PRNT,TAPE,IFL)	RETV0231
300	CONTINUE		RETV0232
C		GO BACK FOR ANOTHER RUN	RETV0233
301	IF(TAPE) ENDFILE IFIL		RETV0234
	WRITE(6,302)		RETV0235
302	FORMAT(////' END OF THIS RUN'/1H1)		RETV0236
	GO TO 1001		RETV0237
	END		RETV0238

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ERTS READING PROGRAM-EREAD,ESKIP,EINIT

PAGE 2

FOR CENTER FOR RESEARCH, INC.

1	LBL	ERTS,ERTS TAPE READING PROGRAM	ERTS0001
2	TTL	ERTS READING PROGRAM-EREAD,ESKIP,EINIT	ERTS000.
3	TTL	FOR CENTER FOR RESEARCH, INC.	ERTS000
4		DATE 9/11/72	ERTS0001
5	*		ERTS000
6	*	CALLING SEQUENCES ARE---	ERTS000
7	*		ERTS000
8	*	TO INITIALIZE--	ERTS000
9	*		ERTS000
10	*	CALL EINIT(NOLS)	ERTS001
11	*		ERTS001
12	*	EINIT INITIALIZES THE ERTS TAPE SO THAT DATA MAY	ERTS001
13	*	BE READ. IT RETURNS THE NUMBER OF WORDS PER	ERTS001
14	*	SCAN LINE IN THE VARIABLE NOLS.	ERTS001
15	*		ERTS001
16	*	TO SKIP A NUMBER OF RECORDS	ERTS001
17	*		ERTS001
18	*	CALL ESKIP(NOSK)	ERTS001
19	*		ERTS002
20	*	NOSK IS THE NUMBER OF RECORDS TO SKIP.	ERTS002
21	*	IF THE END OF FILE IS ENCOUNTERED BEFORE NOSK RECORDS	ERTS002
22	*	ARE SKIPPED, EREAD IS ABORTED.	ERTS002
23	*		ERTS002
24	*	TO READ A LINE	ERTS002
25	*		ERTS002
26	*	CALL EREAD(I,LN)	ERTS002
27	*		ERTS002
28	*	I THIS IS THE ARRAY INTO WHICH THE NOLS WORDS OF DATA	ERTS002
29	*	FROM A LINE OF ERTS DATA IS PLACED. THE DATA IS PLACED INTO	ERTS002
30	*	THIS ARRAY IN STANDARD CORRESPONDING POINT FORM.	ERTS003
31	*	THE ERTS DATA HAS FOUR CHANNELS, SO THERE ARE	ERTS003
32	*	ACTUALLY NOLS/4 DATA POINTS PER SCAN LINE.	ERTS003
33	*	LN THIS IS RETURNED BY EREAD. IF LN IS RETURNED AS ZERO, THE	ERTS003
34	*	END OF FILE WAS REACHED ON THE ERTS TAPE. IF IT IS	ERTS003
35	*	RETURNED NON-ZERO, THEN IT IS THE LINE NUMBER	ERTS003
36	*	OF THE LINE OF DATA RETURNED.	ERTS003
37	*		ERTS003
38	*	NOTE-- THE ERTS TAPE MUST HAVE FILE CODE *ES* FOR	ERTS003
39	*	THIS PROGRAM...	ERTS003
40	*		ERTS004
41	*	ABORT CODES POSSIBLE FROM THIS SUBROUTINE ARE--	ERTS004
42	*	MB EREAD BUFFER 'DATA' IS NOT LARGE ENOUGH	ERTS004
43	*	FOR A BLOCK OF ERTS DATA. IT MUST BE INCREASED	ERTS004
44	*	IN SIZE.	ERTS004
45	*	AI EINIT WAS CALLED TWICE.	ERTS004
46	*	NI EINIT WAS NOT CALLED BEFORE CALLING EREAD OR ESKIP	ERTS004
47	*	UE END OF FILE ENCOUNTERED WHILE READING	ERTS004
48	*	ID OR ANNOTATION BLOCKS ON ERTS TAPE	ERTS004
49	*	EF END OF FILE ENCOUNTERED WHILE TRYING TO SKIP	ERTS004
50	*	RECORDS ON ERTS TAPE IN ESKIP.	ERTS005

STORAGE

53	.EINIT	DFC	0	THIS IS NONZERO IF EINIT HAS BEEN CALLED	ERTS0053
54	RECNO.	DFC	0	THIS IS THE CURRENT LINE NO.	ERTS0054
55	FDW	VFD	18/FCB,1/0,1/1,2/0,1/0,1/0,1/1		ERTS0055
56		FILCB	FCB,ES,,,1024,,,1,,1		ERTS0056
57	CNT1	IOTD	DATA,9	DCW FOR READING ID	ERTS0057
58	CNT2	IOTD	DATA,139	DCW FOR READING ANNOTATION BLOCK	ERTS0058
59	CONT	IOTD	DATA,**	DCW FOR READING DATA BLOCK	ERTS0059
60	EOFA	LDQ	=3H0UE,QL		ERTS0060
61		MME	GEBORT		ERTS0061
62	DATA	BSS	1024	DATA BUFFER AREA	ERTS0062
63	OT	BSS	1	OUTPUT TALLY	ERTS0063
64	IT	BSS	1	INPUT TALLY	ERTS0064
65	STS	TALLY	STK,4	STACK TALLY	ERTS0065
66	ST	BSS	1	STACK TALLY	ERTS0066
67	STK	BSS	4	STACK AREA	ERTS0067

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EINIT

69 EINIT SAVE

ERTS0069

70	SZN	,EINIT	HAVE WE BEEN CALLED BEFORE	ERTS0070
71	TNZ	2ABTA	YES	ERTS0071
72	STC1	,EINIT	MARK THAT WE'VE BEEN HERE ONCE	ERTS0072
73	EAX0	2,1*		ERTS0073
74	STX0	EIP	SAVE ADDRESS	ERTS0074
75	CALL	OPEN(FDW,1)	OPEN FILE	ERTS0075

76	CALL	READ(FCB,CNT1)	READ ID RECORD	ERTS0076
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77	CALL	WAIT(FCB,EOFA)	AND WAIT FOR IT TO GET DONE	ERTS0077
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78	LDQ	DATA+3	GET RECORD LENGTH	ERTS0078
79	ANQ	=0177777,DL	ISOLATE IT	ERTS0079
80	EIP	STQ	**	ERTS0080
81	QLS	6+18	GIVE IT TO CALLER	ERTS0081
82	STCQ	ADD CNT,70	SAVE TALLY COUNT	ERTS0082
83	QRL	6+18-1	MULTIPLY BY 2	ERTS0083
84	DIV	9,DL	AND DIVIDE BY 9	ERTS0084
85	ARL	0	IS REMAINDER ZERO	ERTS0085
86	YZE	*+2	YES	ERTS0086
87	ADQ	1,DL	NO, SO INCREMENT QUOTIENT	ERTS0087
88	CMPO	1025,DL	IS IT TOO BIG	ERTS0088
89	TNC	*+3	NO	ERTS0089

90	LDQ	=3HOMB,DL		ERTS0090
91	MME	GEABRT	YES, SO ABORT	ERTS0091
92	STCQ	CONT,07	SAVE IT IN DCW	ERTS0092
93	MME	GESNAP		ERTS0093
94	ZERO	DATA,9	SNAP OUT ID BLOCK	ERTS0094
95	CALL	READ(FCB,CNT?)	READ ANNOTATION BLOCK	ERTS0095

96	CALL	WAIT(FCB,EOFA)	WAIT ON IT	ERTS0096
----	------	----------------	------------	----------

EINIT

97	MME	GESNAP		ERTS009
98	ZERO	DATA,139	SNAP OUT ANNOATION BLOCK	ERTS009
99	RETURN	EINIT	AND RETURN	ERTS009
100	ESKIP	SAVE		ERTS010

101	SZN	.EINIT	ARE WE INITIALIZED	ERTS010
102	TZE	EABTB	NO	ERTS010
103	LDQ	2,1*	GET NUMBER RECORDS TO SKIP	ERTS010
104	ECOMP	TZE	NONE TO DO, SO RETURN	ERTS010
105	CMPQ	64,DL	IS IT > 63	ERTS010
106	TRC	ESK64	YES	ERTS010
107	ASQ	RECNO.	INCREMENT RECORD COUNT	ERTS010
108	EAX1	0,QL		ERTS010
109	STX1	*+5		ERTS010
110	CALL	FSREC(FCB,.ESKIP,EDFS)		ERTS011

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111	ESKR	RETURN	ESKIP		ERTS011
112	.ESKIP	BSS	1	NO. OF RECORDS TO SKIP	ERTS011
113	EOFS	LDQ	=3H0EF,DL	UNEXPECTED EOF	ERTS011
114		MME	GEBORT		ERTS011
115	* TRYING TO SKIP 64 OR MORE RECORDS				ERTS011
116	ESK64	LDA	63,DL		ERTS011
117		ASA	RECNO.	INCREMENT RECORD COUNT	ERTS011
118		SPQ	63,DL		ERTS011
119		STQ	.ESKIP	SAVE FOR LATER	ERTS011
120		CALL	FSREC(FCB,63,EDFS)		ERTS011

121		LDQ	.ESKIP		ERTS011
122		TRA	ECOMP		ERTS011
123	*				ERTS011
124	EABTA	LDQ	=3H0AI,DL	TRIED TO CALL EINIT TWICE	ERTS011
125		MME	GEBORT		ERTS011
126	EABTB	LDQ	=3H0BI,DL	DIDN'T CALL EINIT	ERTS011



ERTS READING PROGRAM-EREAD,ESKIP,EINIT

PAGE 7

EINIT

ERTS012;

127 MME GEBORT

EREAD

129 EREAD SAVE

ERTS0129

130	SZN	.EINIT	ARE WE INITIALIZED	ERTS0130
131	TZE	EAST3	NO	ERTS0131
132	EAQ	2,1*	GET ARRAY ADDR.	ERTS0132
133	ADDCNT	** ,DL	ADD TALLY COUNT	ERTS0133
134	STQ	OT	AND SAVE IT	ERTS0134
135	EAQ	DATA		ERTS0135
136	STQ	IT	SET UP INPUT TALLY	ERTS0136
137	EAXO	3,1*	INSERT ADDRESS OF COUNT WD	ERTS0137
138	STXO	ERP		ERTS0138
139	CALL	READ(FCB,CONT)	READ NEXT LINE OF DATA	ERTS0139

140	CALL	WAIT(FCB,EOF)	WAIT ON IT	ERTS0140
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141	AOS	RECNO.	INCREMENT LINE NO.	ERTS0141
142	LDQ	RECNO.	GET LINE NO.	ERTS0142
143	ERP	**	RETURN TO CALLER	ERTS0143
144	STZ	LINIT	INITIALIZE EXPANDER	ERTS0144
145	NEXT	STS		ERTS0145
146	LDQ	ST	INITIZE STACK TALLY	ERTS0146
147	NXTONE	L	GET POINT	ERTS0147
148	STA	OT, ID	AND PUT IT IN ARRAY	ERTS0148
149	TTF	**2		ERTS0149
150	DRL		ERROR	ERTS0150
151	TSX1	L	GET NEXT POINT	ERTS0151
152	STA	ST, ID	PUT INTO STACK TALLY	ERTS0152
153	TTF	NXTONE	GO PROCESS NEXT	ERTS0153
154	LDQ	STS	TALLY RUNOUT	ERTS0154
155	STQ	ST	SO REINITIALIZE TALLY	ERTS0155
156	AGAIN	ST, ID	PICK UP POINTS	ERTS0156
157	TTF	MORE	GOT ONE	ERTS0157
158	STQ	OT, ID	TALLY RUNOUT	ERTS0158
159	TTF	NEXT	GO PROCESS NEXT EIGHT POINTS.	ERTS0159
160	RETURN	EREAD	TALLY RUNOUT, SO WE'RE DONE.	ERTS0160
161	MORE	OT, ID	SAVE IT	ERTS0161
162	TTF	AGAIN	GO GET NEXT ONE STACKED	ERTS0162
163	DRL		ERROR	ERTS0163

POINT FETCHING ROUTINE L

165	LSAVE	BSS	1	QR SAVED FROM LAST TSX1 L	ERTS0165
166	LINIT	BSS	1	SHIFT INDICATOR FOR L	ERTS0166
167	*				ERTS0167
168	L	SZN	LINIT	ARE WE READY FOR NEXT PAIR	ERTS0168
169		TNZ	LCONT	YES	ERTS0169
170		LDQ	IT,IO	NO, SO GET NEXT WORD	ERTS0170
171		AOS	LINIT	INCREMENT LINIT	ERTS0171
172		TRA	LRET	DONE FOR THIS ONE	ERTS0172
173	LCONT	AOS	LINIT	INCREMENT AGAIN.	ERTS0173
174		LDQ	LINIT		ERTS0174
175		CMPQ	5,DL	IS THIS 5TH TIME	ERTS0175
176		TZF	L5	YES	ERTS0176
177		CMPQ	9,DL	OR 9TH TIME	ERTS0177
178		TNZ	*+2	NO	ERTS0178
179		STZ	LINIT	REINITIALIZE	ERTS0179
180		LDQ	LSAVE	GET SAVED QR	ERTS0180
181	LRET	LDA	0,DL		ERTS0181
182		LLS	8	SHIFT OVER	ERTS0182
183		STQ	LSAVE	SAVE QR FOR NEXT TIME	ERTS0183
184		TRA	0,1	AND RETURN	ERTS0184
185	L5	LDA	0,DL		ERTS0185
186		LDQ	LSAVE	RESTORE QR	ERTS0186
187		LLS	4	THIS ONE IS DIVIDED OVER	ERTS0187
188		LDQ	IT,IO	WORD BOUNDARIES	ERTS0188
189		LLS	4		ERTS0189
190		STQ	LSAVE		ERTS0190
191		TRA	0,1	RETURN	ERTS0191
192	*				ERTS0192
193	* EOF	PROCESSOR FOR EREAD			ERTS0193
194	EOF	STZ	ERP,I	MARK EOF TO CALLER	ERTS0194
195		RETURN	EREAD	AND RETURN	ERTS0195

196            END  
 TION.  
 PA 091572/091472    JMPB 053172/070872    JMPC 072772/072772  
 IN THE ABOVE ASSEMBLY

ERTS0196

CPIXFY

## SPECIAL PICTURE ROUTINE

JAN 1974

WRITTEN BY RJ BOSLEY

## DESCRIPTION OF PROGRAM.

THIS SUBPROGRAM OUTPUTS A PICTURE OF THE ERTS DATA FROM USER SPECIFIED PARAMETERS TO PITCHR. INPUT PARAMETERS FOR PITCHR PLUS THE PARAMETERS LNSKIP AND KOLSKP ALL MUST BE SUPPLIED TO THIS SUBPROGRAM UNDER THE NAMELIST \$PICTUR FORMAT. FOR A COMPLETE DESCRIPTION OF INPUT ARGUMENTS TO PITCHR, SEE THE PITCHR WRITEUP. BY THE USER SETTING IMAX=0 AND IMIN= MAX GREY TONE LEVEL, IT IS POSSIBLE TO OBTAIN A POSITIVE IMAGE INSTEAD OF A NEGATIVE.

NOTE--DISC FILES MUST BE ON FILE CODES 13 AND 14.

## INTERNAL PARAMETERS.

IBAND  
IS,IE  
LINES  
NOSK  
IPSTR,IPEND  
LNSKIP,KOLSKP  
ISKIP  
KP

BAND NUMBER BEING PROCESSED  
STARTING, ENDING BAND NUMBERS  
NUMBER OF LINES IN IMAGE PRINTED  
NUMBER OF LINES TO SKIP IN ERTS TAPE  
STARTING, ENDING POINTS IN ERTS LINE  
LINE AND COLUMN INCREMENT  
COLUMN INCREMENT FOR ERTS LINE  
NUMBER OF COLUMNS IN IMAGE

## INPUT ARGUMENTS TO PITCHR.

IMAGE  
ICELL, JCELL  
INIT

IMIN, IMAX  
NROW  
NFILES

IFIL  
NULW, NULD  
AMAG, DMAG

ARRAY TO BE PRINTED  
NUMBER OF ROWS, COLS IN ARRAY  
NUMBER OF TIMES ENTRY IS MADE AT SNAP  
MUST BE GREATER THAN 1  
MINIMUM, MAXIMUM GREY TONES IN ARRAY  
NUMBER OF ROWS TO BE PRINTED=ICELL  
NUMBER OF OUTPUT FILES AVAILABLE,  
SET=0 FOR ALL OUTPUT ON FILE CODE 6  
SET=2 FOR FILES 06 AND 42  
IF .GT.2, USER MUST SUPPLY FILES  
ARRAY CONTAINING OUTPUT FILE CODES  
NUMBER OF COLS, ROWS PER OUTPUT PAGE  
WIDTH, LENGTH DOWN MAGNIFICATION

## ENTRY POINT.

CALL PIXEY(ILINE, IMAGE, IRSTRT, IRSTOP, ICSTRT, ICSTOP, QUAN, NBAND)

## INPUT ARGUMENTS.

ILINE  
IMAGE  
IRSTRT, IRSTOP  
ICSTRT, ICSTOP  
QUAN

NBAND

ARRAY ERTS LINE IS READ INTO  
IMAGE ARRAY FOR PITCHR, HOLDS ONE LINE  
STARTING, STOPPING ROW IN IMAGE  
STARTING, STOPPING COL IN IMAGE  
TRUE FOR EQUAL PROBABILITY QUANTIZING  
OF IMAGE  
BAND NUMBER TO BE PROCESSED

PIXEY001  
PIXEY002  
PIXEY003  
PIXEY004  
PIXEY005  
PIXEY006  
PIXEY007  
PIXEY008  
PIXEY009  
PIXEY010  
PIXEY011  
PIXEY012  
PIXEY013  
PIXEY014  
PIXEY015  
PIXEY016  
PIXEY017  
PIXEY018  
PIXEY019  
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PIXEY034  
PIXEY035  
PIXEY036  
PIXEY037  
PIXEY038  
PIXEY039  
PIXEY040  
PIXEY041  
PIXEY042  
PIXEY043  
PIXEY044  
PIXEY045  
PIXEY046  
PIXEY047  
PIXEY048  
PIXEY049  
PIXEY050

```

C                                     SET=5 FOR ALL BANDS
C
C                                     PIXFY051
C                                     PIXFY052
C                                     PIXFY053
C                                     PIXFY054
C                                     PIXFY055
C                                     PIXFY056
C                                     PIXFY057
C                                     PIXFY059
C                                     PIXFY059
C                                     PIXFY059
C                                     PIXFY060
C                                     PIXFY061
C                                     PIXFY062
C                                     PIXFY063
C                                     PIXFY064
C                                     PIXFY065
C                                     PIXFY066
C                                     PIXFY067
C                                     PIXFY068
C                                     PIXFY069
C                                     PIXFY070
C                                     PIXFY071
C                                     PIXFY072
C                                     PIXFY073
C                                     PIXFY074
C                                     PIXFY075
C                                     PIXFY076
C                                     PIXFY077
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C                                     PIXFY079
C                                     PIXFY080
C                                     PIXFY081
C                                     PIXFY082
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C                                     PIXFY093
C                                     PIXFY094
C                                     PIXFY095
C                                     PIXFY096
C                                     PIXFY097
C                                     PIXFY098
C                                     PIXFY099
C                                     PIXFY100
C
C     EXAMPLE OF SAMPLE RUN.
C     CONTROL CARDS--
C     $PARAM SPIC=T,QUAN=T,IRSTRT=1,IRSTOP=1216,ICSTRT=5,
C     ICSTOP=772,NRAND=23END
C     $PICTUR INIT=304,NFILES=2,LNSKIP=4,KOLSKP=3,JCELL=256$END
C     THIS RUN WILL PRINT OUT ON FILES 06 AND 42 A PICTURE 256 PTS
C     WIDE BY 304 LINES. NOTE THAT 1216/4=304 AND 768/3=256, GIVING THE
C     VALUES FOR LNSKIP=4,KOLSKP=3,JCELL=256,INIT=304. ALSO, USING
C     THESE VALUES FOR LNSKIP AND KOLSKP, THE PICTURE WILL BE IN
C     PROPORTION TO THE ERTS PRINT.
C
C     SUBROUTINE PIXEY(ILINE,IMAGE,IRSTRT,IPSTOP,ICSTRT,ICSTOP,QUAN,
1     NBAND)
C
C     DIMENSION ILINE(1),IFIL(10),IMAGE(1)
C     LOGICAL QUAN
C     NAMELIST /PICTUR/NFILES,IFIL,ICELL,JCELL,INIT,IMIN,IMAX,NROW,
1     NULW,NULD,AMAG,DMAG,LNSKIP,KOLSKP
C     INITIALIZE PARAMETERS
C
C     ICELL=1
C     JCELL=256
C     INIT=304
C     IMIN=0
C     IMAX=75
C     IF(QUAN) IMAX=12
C     NFILES=?
C     IFIL(1)=6
C     IFIL(2)=42
C     NROW=ICELL
C     NULW=129
C     NULD=60
C     AMAG=1
C     DMAG=1
C     LNSKIP=4
C     KOLSKP=3
C
C     READ PARAMETERS
C
C     READ(5,PICTUR)
C
C     WRITE(6,1)
C     WRITE(6,PICTUR)
C     WRITE OUT PARAMETERS
1     FORMAT(////20X,'SPECIAL PICTURE ROUTINE'////)
2     FORMAT(1H1,'BAND NUMBER ',I1)
C     SET UP FOR THE DESIRED BAND
C
C     IS=1
C     IF=4
C     ISKIP=4*KOLSKP
C     IF(NBAND.EQ.5) GO TO 5

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REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

02-09-74 20.623

SPECIAL PICTURE ROUTINE

```
IS=NPAND
IE=IS
C          GO THRU EACH BAND
5 DO 100 IBAND=IS,IE
  REWIND 13
  REWIND 14
C          WRITE SUBTITLE
  WRITE(6,2) IBAND
  IPSTR=(ICSTPT-1)*4+IBAND
  IPEND=ICSTOP*4
C          INITIALIZE SNAP IN PITCHR
  CALL PITCHR(IMAGE,ICELL,JCELL,INIT,1,IMAX,IMIN,NROW,NFILES,IFIL,
1          NULW,NULD,AMAG,DMAG,)
C          POSITION THE INPUT TAPE
  NOSK=IRSTRT-1
  CALL EFWND
  IF(NOSK.NE.0) CALL ESKIP(NOSK)
C          GO THRU EACH LINE
  DO 50 LINE=IPSTR,IRSTOP,LNSKIP
C          READ LNSKIP LINES
  DO 10 LIN=1,LNSKIP
  CALL EREAD(ILINE,LN)
  IF(LN.NE.0) GO TO 10
  L=LINE+LIN
  WRITE(6,9) L
9  FORMAT(//// * **EXECUTION TERMINATED--EOF DETECTED ON ERTS TAPE,
1LINE NUMBER IS *,I6/IH1)
  RETURN
10 CONTINUE
C          GET THIS LINE FOR THE IMAGE
  KP=0
  DO 20 IP=IPSTR,IPEND,ISKIP
  KP=KP+1
  IMAGE(KP)=ILINE(IP)
20 CONTINUE
C          WRITE THIS LINE TO SCRATCH FILE 14
  WRITE(14) (IMAGE(IP),IP=1,JCELL)
50 CONTINUE
  ENDFILE 14
  REWIND 14
C          QUANTIZE FILE 14 AND PUT OUT ON 13
  IF(QUAN) CALL ZEQUAN(IMAGE,INIT,JCELL,1,1,1,13,14,13)
  REWIND 13
C          READ THE QUANTIZED IMAGE
  DO 90 I=1,INIT
  READ(13) (IMAGE(K),K=1,JCELL)
C          SNAP OUT THIS LINE
90 CALL SNAP
100 CONTINUE
  RETURN
  END
```

PIXFY101  
PIXFY102  
PIXFY103  
PIXFY104  
PIXFY105  
PIXFY106  
PIXFY107  
PIXFY108  
PIXFY109  
PIXFY110  
PIXFY111  
PIXFY112  
PIXFY113  
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PIXFY140  
PIXFY141  
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PIXFY143  
PIXFY144  
PIXFY145  
PIXFY146  
PIXFY147  
PIXFY148  
PIXFY149  
PIXFY150

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Z E Q U A N

CZEQUAN

Z E Q U A N

SEPT 1973

MODIFIED FROM KEQUAN BY Z DINSTEN

DESCRIPTION OF PROGRAM.

THIS SUBROUTINE QUANTIZES AN IMAGE ON FILE 'INFILE' BY EQUAL PROBABILITY TO NQ LEVELS AND OUTPUTS IT TO FILE 'IOUTFL'. THIS PROCESSING IS DONE LINE BY LINE AFTER A FIRST PASS IS MADE THRU THE IMAGE TO DETERMINE THE MINIMUM AND MAXIMUM GREY TONES.

ENTRY POINT.

CALL ZEQUAN(LINE, NUMLIN, NUMPPL, NCOMP, ICOMP, LEFT, NQ, INFILE, IOUTFL)

INPUT ARGUMENTS.

LINE = ARRAY ONE LINE OF IMAGE IS READ INTO  
NUMLIN= NUMBER OF LINES IN IMAGE  
NUMPPL= NUMBER OF COLUMNS IN IMAGE  
NCOMP = NUMBER OF COMPONENTS IN ORIGINAL IMAGE  
ICOMP = THE COMPONENT TO BE QUANTIZED  
LEFT = LEFT-MOST CELL IN LINE DESIRED  
NQ = NUMBER OF QUANTIZED LEVELS  
INFILE= INPUT FILE CONTAINING IMAGE TO BE QUANTIZED

OUTPUT ARGUMENTS.

IOUTFL= OUTPUT FILE CONTAINING QUANTIZED IMAGE

SUBROUTINE ZEQUAN(LINE, NUMLIN, NUMPPL, NCOMP, ICOMP, LEFT, NQ, INFILE, IOUTFL)

DIMENSION LINE(NCOMP, 1), KN(512)

INITIALIZE HISTOGRAM TO ZERO

DO 1 I=1, 512

KN(I)=0

CONTINUE

MIN=10000

MAX=-10000

REWIND INFILE

GO THRU EACH LINE

DO 2 II=i, NUMLIN

READ(INFILE) ((LINE(J, L), J=1, NCOMP), L=1, NUMPPL)

SET STARTING, ENDING PTS IN LINE

IRIGHT=LEFT+NUMPPL-1

DO 2 I=LEFT, IRIGHT

J=LINE(ICOMP, I)

GET MIN, MAX AND HISTOGRAM

IF (MAX.LT.J) MAX=J

IF (MIN.GT.J) MIN=J

KN(J+1)=KN(J+1)+1

ZEQUAN01  
ZEQUAN02  
ZEQUAN03  
ZEQUAN04  
ZEQUAN05  
ZEQUAN06  
ZEQUAN07  
ZEQUAN08  
ZEQUAN09  
ZEQUAN10  
ZEQUAN11  
ZEQUAN12  
ZEQUAN13  
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ZEQUAN47  
ZEQUAN48  
ZEQUAN49  
ZEQUAN50

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Z E Q U A N

2	CONTINUE		ZEQUAN51
	NGL=MAX		ZEQUAN52
	WRITE(6,100) MIN,MAX		ZFUQUAN53
100	FORMAT(1X,'MIN, MAX APE ',2I4)		ZEQUAN54
C		GET NO OF PTS IN IMAGE	ZEQUAN55
	NP=NUMLIN*NUMPPL		ZEQUAN56
	J=1		ZEQUAN57
	MQ=NQ		ZEQUAN58
3		DETERMINE LEVELS	ZEQUAN59
	DO 3 I=1,NQ		ZEQUAN60
	NL=NP		ZFUQUAN61
C		GET NEW LEVEL	ZEQUAN62
4	NL=NL-MQ*KN(J)		ZEQUAN63
	NP=NP-KN(J)		ZEQUAN64
	KN(J)=I-1		ZEQUAN65
	J=J+1		ZEQUAN66
C		IF LAST LEVEL,SKIP	ZEQUAN67
	IF(J.GT.NGL) GO TO 6		ZEQUAN68
C		INCREMENT THE LEVEL AGAIN	ZEQUAN69
	IF(MQ*KN(J).LE.NL*2)GO TO 4		ZEQUAN70
C		DECREASE THE NO. OF LEVELS LEFT	ZEQUAN71
3	MQ=MQ-1		ZEQUAN72
	DO 5 I=J,NGL		ZEQUAN73
5	KN(I)=MQ-1		ZEQUAN74
	GO TO 8		ZEQUAN75
C		RESET THE LAST LEVEL	ZEQUAN76
6	N=(NQ-I)/2		ZEQUAN77
	IF(N.LT.1)GO TO 8		ZEQUAN78
	DO 7 I=1,NGL		ZEQUAN79
7	KN(I)=KN(I)+N		ZEQUAN80
8	CONTINUE		ZEQUAN81
C		REWIND DISC FILES	ZEQUAN82
	REWIND INFILE		ZEQUAN83
	REWIND IOUTFL		ZEQUAN84
3		ASSIGN QUANTIZED LEVELS LINE BY LINE	ZEQUAN85
	DO 11 II=1,NUMLIN		ZEQUAN86
	READ(INFILE) ((LINE(J,L),J=1,NCOMP),L=1,NUMPPL)		ZEQUAN87
	DO 9 I=LEFT,IRIGHT		ZEQUAN88
	J=LINE(ICOMP,I)		ZEQUAN89
9	LINE(ICOMP,I)=KN(J+1)		ZEQUAN90
C		WRITE OUT QUANTIZED FILE	ZEQUAN91
	WRITE(IOUTFL) (LINE(ICOMP,K),K=LEFT,IRIGHT)		ZEQUAN92
11	CONTINUE		ZEQUAN93
	ENDFILE IOUTFL		ZEQUAN94
	REWIND IOUTFL		ZEQUAN95
	RETURN		ZEQUAN96
	END		ZEQUAN97



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PTCHRG TYPE EXECUTION REPORT

```

5  GMAP  WPECK,COMPX  P I T C H R  G M A P
1  TTL  P I T C H R  P I T C H R  G M A P
2  LBL  PTCHRG,
3  REM  ENTRY POINTS ARGSET & MARG BY G ELLIOTT,
4  REM
5  SYMDEF ARGSET
6  ARGSET SAVE
7  LDD  1,DL
8  MVE  GESETS
9  LDX0 1,1
10 LDX0 0,0
11 EAX0 1,0
12 STX0 LOCAT
13 LDX0 -1,0
14 SELX0 LOCAT
15 STX0 MAXNUM
16 RETURN ARGSET
17 SYMDEF MARG
18 MARG SAVE
19 LXLO 2,1*
20 EAO 0
21 MAXNUM C*P*Y*O **,*DU
22 TMI **+3
23 SZY 0,*DU
24 TRA ,E,L,,,1
25 LOCAT EAX0 **,0
26 LDX0 U,0
27 TVZ **+2
28 TRA ,E,L,,,1
29 LDD 1,DL
30 TRA ,E,L,,,1
31 SYMDEF APORT
32 E BSS 1
33 ABORT NULL
34 LDD =3NOPT,DL
35 MVE GEBORT
36 SYMDEF BORED
37 BORED EAX0 3,1*
38 LDD 4,1*
39 OLS 6
40 STO TALI
41 STX0 TALI
42 EAX0 =6H-
43 STX0 STR-1
44 EAX0 TW
45 STX0 STR
46 TRA BOR
47 TT ADS 2,1*
48 TTF BCR
49 TRA 0,1*
50 TALI BSS 1
51 SYMDEF BORDER

```

TO SET BIT 35 & INITIALIZE FOR MARG, SET BIT 35

TRACE BACK TO GET LOCATION OF LAST CALL TO PITCHR PUT IT WHERE MARG CAN USE IT, GET NUMBER OF ARGUMENTS PROVIDED, ALSO FOR MARG;

ENTRY POINT TO CHECK FOR THE EXISTENCE OF NON-NULL ARGUMENTS, WHICH ARGUMENT AREA CHECKING ?

DO WE HAVE THAT MANY ?

NO, GO HOME!

IS IT A NULL POINTER ?

YES  
NO

GET ADDRESS OF ARRAY, NO. ELEMENTS IN ARRAY, USE FOR TALLY COUNT

CHARACTER THAT GOES INTO ARRAY, XED WILL USE TALLY MODIFICATION

START FILLING ARRAY, BUMP BORDER POSITION

PTCHRG01  
PTCHRG02  
PTCHRG03  
PTCHRG04  
PTCHRG05  
PTCHRG06  
PTCHRG07  
PTCHRG08  
PTCHRG09  
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PTCHRG48  
PTCHRG49  
PTCHRG50  
PTCHRG51

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PTCHRG INDEX EXECUTION REPORT

52	BORDER	EAX1	00
53		STX0	STR
54		EAX0	=5H1
55		STX0	STR-1
56	FOR	LDD	2,1*
57		DIV	10,DL
58		CMPA	0,DL
59		TNZ	1,TEN
60		DIV	10,DL
61		LDD	HLNK
62		LLS	30
63		TRA	STR
64	NTEN	CMPA	5,DL
65		TZE	FIVE
66		LDA	**
67	STR	XED	**
68		TRA	0,1
69	FIVE	LDA	=6H*
70		TRA	STR
71	TH	ESTA	TALI, ID
72		TRA	TT
73	DS	STA	3,1*
74		STA	4,1*
75		SYMDEF	LSHIFT
76	LSHIFT	EAX0	2,1*
77		LDD	3,1*
78		QLS	6
79		STD	TALI
80		STX0	TALI
81	LD	LDD	TALI, 1
82		QLS	6
83		STD	TALI, ID
84		TF	LD
85		TRA	0,1*
86		SYMDEF	IFETCH
87	S1	BSS	1
88	S2	BSS	1
89	S3	BSS	1
90	IFETCH	SAVE	2,7
91		LXL7	4,1*
92		EAX2	2,1*
93		TSX0	GETT
94	SS1	FSTR	S1
95		LXL0	7,1*
96		TNZ	*+3
97		ADX2	1,DU
98		TSX0	GETT
99		FSTR	S2
100		LXL0	0,1*
101		TZE	*+3
102		FLD	S1
103		TRA	FS

WE STORE ON EITHER SIDE OF IMAGE DATA CHARACTER FOR VERTICAL BORDER

CURRENT LINE COUNT IS IT DIVISIBLE BY TEN

PUT BLANKS IN OR MAKE NO. LEFT JUSTIFIED, BLANK FILLED

WAS THE NUMBER DIVISIBLE BY FIVE

RETURN IF WE ENTERED AT BORDER

PUT IN HOR BORDER ARRAY

ADDRESS OF ARRAY NO. ELEMENTS IN ARRAY USE FOR TALLY COUNT

SHIFT UP NEXT CHARACTER

=1 FOR INTEGER, 0 FOR FLOATING PT

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PTCHRG52  
PTCHRG53  
PTCHRG54  
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PTCHRG61  
PTCHRG62  
PTCHRG63  
PTCHRG64  
PTCHRG65  
PTCHRG66  
PTCHRG67  
PTCHRG68  
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PTCHRG71  
PTCHRG72  
PTCHRG73  
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PTCHRG90  
PTCHRG91  
PTCHRG92  
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PTCHRG97  
PTCHRG98  
PTCHRG99  
PTCHRG00  
PTCHRG01  
PTCHRG02  
PTCHRG03

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PTCHRG

CMPEX EXECUTION REPORT

104		LYL2	3,1*
105		ADX2	2,1
106		YSA	GETT
107		FSS	S1
108		F4P	6,1*
109		FAD	S1
110	FS	FST3	S3
111		FLD	S2
112		FSS	S1
113		F4P	5,1*
114		FAD	S1
115		FAD	S3
116		FDV	=2,
117	FM	FSS	FMIN
118	IE	EDU	FM-SS1
119		FDV	FDIF
120		F4P	E
121		FAD	=0,,DU
122		TPL	**2
123		UFA	=0777777777
124		JFA	71*1024,DU
125		ADG	1,DL
126		CYPO	2,DL
127		TPL	**3
128		LDD	1,DL
129		RETURN	IFETCH
130	ID	CYPO	**DL
131		IMI	**2
132	IDD	LDD	**DL
133		RETURN	IFETCH
134	GETT	CMPX7	1,DU
135		TZE	3,IC
136		FLD	0,2
137	01	TRA	**0
138		LDA	0,2
139		EAG	0
140		LDE	35*1024,DU
141		FAD	=0,,DU
142	02	TRA	**0
143		SYNDEF	FI
144	F1	FLD	2,1*
145		FSS	3,1*
146		FST	FDIF
147		FLD	3,1*
148		FST	FMIN
149		LXL0	4,1*
150		STX0	IDD
151		ADX1	1,DU
152		STX0	ID
153		EAA	0
154		EAG	-1,0
155		LLS	10

PTCHRG04  
PTCHRG05  
PTCHRG06  
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PTCHRG54  
PTCHRG55

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PTCHPG

INDEX EXECUTION REPORT

156		LDE	35*1024,00
157		FAD	=0,,DU
158		FSTR	E
159		LDD	0,2
160		LUS	36
161		LXLQ	5,1*
162		TNZ	ZZ
163		EAXQ	IE
164	SX	STXQ	01
165		STXQ	02
166		TRA	0,1*
167	ZZ	EAXQ	0
168		TRA	SX
169	FDIF	BSS	1
170	FMIN	BSS	1
171	BLVK	BCI	1,
172		EVD	

PTCHRG56  
PTCHRG57  
PTCHRG58  
PTCHRG59  
PTCHRG60  
PTCHRG61  
PTCHRG62  
PTCHRG63  
PTCHRG64  
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PTCHRG66  
PTCHRG67  
PTCHRG68  
PTCHRG69  
PTCHRG70  
PTCHRG71  
PTCHRG72

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WRITE ERTS DATA ONTO D.S.C

```

CWRDASK          WRITE ERTS DATA ONTO DISC
C
C              VERSION 17 WRITTEN BY RJ LOSLEY          JAN 1974
C
C              DESCRIPTION OF PROGRAM.
C              THIS SUBROUTINE WRITES ONE LINE OF ERTS DATA FROM ARRAY ILINE
C              ONTO DISC FILES, ONE FILE FOR EACH BAND. DATA IS WRITTEN IN NHOR
C              BLOCKS.
C
C              INTERNAL PARAMETERS.
C              NDSK          DISC FILE CODE, 11 THRU 14
C              IP           POINT INDEX FOR ERTS LINE
C              IPSTR,IPSTOP STARTING,STOPING POINTS IN LINE,
C                          DEPENDENT UPON THE BAND
C
C              ENTRY POINT.
C              CALL WRDASK(ILINE,NHOR,IPSTR,IPEND,NBAND)
C
C              INPUT ARGUMENTS.
C              ILINE        ARRAY CONTAINING ERTS LINE OF DATA
C              NHOR         NUMBER OF HORIZONTAL BLOCKS OF 41 COL
C              IPSTR,IPEND  START AND END POINTS IN EPTS LINE
C              NBAND        THE BAND DESIRED, SET=5 FOR ALL FOUR
C
C              SUBROUTINE WRDASK(ILINE,NHOR,IPSTR,IPEND,NBAND)
C
C              DIMENSION ILINE(1)
C              IF (NBAND.EQ.5) GO TO 100
C              IF ALL BANDS,SKIP DOWN
C              WRITE ONE BAND ONLY
C              NDSK=10+NBAND
C              WRITE NHOR BLOCKS
C              DO 10 J=1,NHOR
C              IPSTR=(J-1)*41*4+NBAND+IPSTR
C              IPSTOP=IPSTR+40*4
C              WRITE(NDSK) (ILINE(IP),IP=IPSTR,IPSTOP,4)
C              CONTINUE
C              RETURN
C              CONTINUE
C              WRITE ALL FOUR BANDS
C              DO 101 I=1,4
C              NDSK=10+I
C              WRITE NHOR BLOCKS
C              DO 101 J=1,NHOR
C              IPSTR=(J-1)*41*4+I+IPSTR
C              IPSTOP=IPSTR+40*4
C              WRITE(NDSK) (ILINE(IP),IP=IPSTR,IPSTOP,4)
C              CONTINUE
C              RETURN
C              END
C
CWRDASK01
CWRDASK02
CWRDASK03
CWRDASK04
CWRDASK05
CWRDASK06
CWRDASK07
CWRDASK08
CWRDASK09
CWRDASK10
CWRDASK11
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CWRDASK49
CWRDASK50

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C      CRDSK1      READ DISC AND OUTPUT PICTURE      RDCSK101
C      VERSION I. WRITTEN BY RJ BOSLEY      JAN 1974      RDCSK102
C      DESCRIPTION OF PROGRAM.      RDCSK103
C      THIS SUBROUTINE READS THE ERTS DATA FROM THE DISC FILES AND      RDCSK104
C      THEN PRINTS OUT A PICTURE OF THE DATA, PROCEEDING VERTICALLY AND      RDCSK105
C      THEN FROM LEFT TO RIGHT. SINCE THE BLOCKS ARE 41 COLUMNS THEN TWO      RDCSK106
C      WILL FIT ON ONE PAGE, ELIMINATING ONE HALF OF THE TOTAL LINES.      RDCSK107
C      INTERNAL PARAMETERS.      RDCSK108
C      NGL      NUMBER OF GREY LEVELS IN IMAGE      RDCSK109
C      NQ      NUMBER OF QUANTIZING LEVELS      RDCSK110
C      IASIZE      SIZE OF ARRAY TO BE QUANTIZED      RDCSK111
C      MHOR      HALF OF NHOR      RDCSK112
C      NCOL      NUMBER OF COLUMNS USED IN IMAGE      RDCSK113
C      NSKIP      NUMBER OF RECORDS TO SKIP TO STAY IN      RDCSK114
C      THE SAME STRIP      RDCSK115
C      LAST      TRUE INDICATES THE LAST STRIP      RDCSK116
C      ENTRY POINT.      RDCSK117
C      CALL RDCSK1 (IMAGE, QUAN, NHOR, NVERT, NDSK)      RDCSK118
C      INPUT ARGUMENTS.      RDCSK119
C      IMAGE      ARRAY TO STORE TWO 41 X 41 BLOCKS      RDCSK120
C      QUAN      TRUE FOR EQUAL PROBABILITY QUANTI-      RDCSK121
C      ZATION OF THE IMAGE      RDCSK122
C      MAX      MAXIMUM GREY TONE IN IMAGE      RDCSK123
C      MIN      MINIMUM GREY TONE IN IMAGE      RDCSK124
C      NHOR      NUMBER OF HORIZONTAL BLOCKS      RDCSK125
C      NVERT      NUMBER OF VERTICAL BLOCKS      RDCSK126
C      NDSK      FILE CODE OF DISC TO BE PROCESSED      RDCSK127
C      SUBROUTINE RDCSK1 (IMAGE, QUAN, NHOR, NVERT, NDSK)      RDCSK128
C      DIMENSION IMAGE (41, 82)      RDCSK129
C      LOGICAL EOF, QUAN, LAST      RDCSK130
C      EOF = .FALSE.      RDCSK131
C      LAST = .FALSE.      RDCSK132
C      CALL FLGEOF (NDSK, EOF)      RDCSK133
C      DETERMINE THE NUMBER OF LEVELS      RDCSK134
C      MAX = 63      RDCSK135
C      MIN = 0      RDCSK136
C      IF (QUAN) MAX = 11      RDCSK137
C      SET UP FOR QUANTIZATION      RDCSK138
C      NGL = 75      RDCSK139
C      NQ = 12      RDCSK140
C      MHOR = NHOR / 2      RDCSK141
C      NCOL = 82      RDCSK142
C      RDCSK143
C      RDCSK144
C      RDCSK145
C      RDCSK146
C      RDCSK147
C      RDCSK148
C      RDCSK149
C      RDCSK150

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READ DISC AND OUTPUT PICTURE

```
3 IF(NHOR.NE.1) GO TO 2
  NCOL=41
  MHOR=1
2 IASIZE=41*NCOL
  C SET THE NUMBER OF RECORDS TO SKIP
  C TO REMAIN IN THE SAME STRIP
  C
  C NSKIP=MHOR-(NCOL/41)
  C GO THRU HORIZONTALLY
  DO 100 J=1,MHOR
  REWIND NDSK
  K=J
  IF(LAST) K=(NHOR/2)+1
  C IF FIRST BLOCK, DO NOT SKIP OVER
  IF(K.EQ.1) GO TO 5
  DO 1 I=2,K
  IF(NHOR.GT.1) READ(NDSK)
  READ(NDSK)
  C IF ONLY ONE BLOCK VERTICALLY, DO NOT
  C INITIALIZE SNAP IN PITCHR
  IF(INVERT.NE.1)
  1 CALL PITCHR(IMAGE,41,NCOL,NVERT,,MAX,MIN,,,,,1.21,,)
  C READ IN TWO 41 BY 41 BLOCKS
  DO 50 II=1,NVERT
  C READ 41 ROWS
  DO 20 JJ=1,41
  C READ NCOL COLUMNS
  READ(NDSK) (IMAGE(JJ,KOL),KOL=1,41)
  IF(NCOL.EQ.82) READ(NDSK) (IMAGE(JJ,KOL),KOL=42,82)
  C SKIP OVER RECORDS NOT WANTED
  IF(NSKIP.EQ.0) GO TO 20
  DO 16 N=1,NSKIP
  READ(NDSK)
  IF(EOF) GO TO 20
  16 CONTINUE
  20 CONTINUE
  EOF=.FALSE.
  C QUANTIZE THE IMAGE
  C IF(QUAN) CALL KEQUAN(IMAGE,NGL,NQ,IASIZE)
  C SNAP OUT A 41 BY 82 BLOCK
  IF(INVERT.NE.1) CALL SNAP
  IF(INVERT.EQ.1) CALL PITCHR(IMAGE,41,NCOL,1,,MAX,MIN,,,,,1.21,,)
  50 CONTINUE
  100 CONTINUE
  C CHECK TO SEE IF NHOR IS EVEN
  IF(NHOR.EQ.1) GO TO 101
  IF(NHOR.EQ.(MHOR*2)) GO TO 101
  C IF NOT, DO THE LAST STRIP OF 41 COLS
  IF(LAST) GO TO 101
  LAST=.TRUE.
  GO TO 3
```

RDCSK151  
RDCSK152  
RDCSK153  
RDCSK154  
RDCSK155  
RDCSK156  
RDCSK157  
RDCSK158  
RDCSK159  
RDCSK160  
RDCSK161  
RDCSK162  
RDCSK163  
RDCSK164  
RDCSK165  
RDCSK166  
RDCSK167  
RDCSK168  
RDCSK169  
RDCSK170  
RDCSK171  
RDCSK172  
RDCSK173  
RDCSK174  
RDCSK175  
RDCSK176  
RDCSK177  
RDCSK178  
RDCSK179  
RDCSK180  
RDCSK181  
RDCSK182  
RDCSK183  
RDCSK184  
RDCSK185  
RDCSK186  
RDCSK187  
RDCSK188  
RDCSK189  
RDCSK190  
RDCSK191  
RDCSK192  
RDCSK193  
RDCSK194  
RDCSK195  
RDCSK196  
RDCSK197  
RDCSK198  
RDCSK199  
RDCSK200

02-12-74 19.078

READ DISC AND OUTPUT PICTURE

```
C.
101 IBAND=NDSK-10          WRITE THE BAND NUMBER
    WRITE(6,102) IBAND
102  FORMAT(6X,'PICTURE FROM BAND NUMBER',I2)
    RETURN
    END
```

R0DSK10  
R0DSK10  
R0DSK10  
R0DSK10  
R0DSK10  
R0DSK10



C KEQUAN

K-E-Q-U-A-N

KEQUAN01

C

WRITTEN BY G. ELLIOT  
VERSION II BY RJ BOSLEY FOR IASIZE

SEPT 1971  
JUNE 1973

KEQUAN02

KEQUAN03

KEQUAN04

KEQUAN05

KEQUAN06

C

DESCRIPTION OF PROGRAM.  
THIS PROGRAM WILL QUANTIZE BY EQUAL PROBABILITY THE INPUT  
ARRAY IA TO NQ LEVELS.

KEQUAN07

KEQUAN08

KEQUAN09

KEQUAN10

C

ENTRY POINT.  
CALL KEQUAN(IA,NGL,NQ,IASIZE)

KEQUAN11

KEQUAN12

KEQUAN13

C

ARGUMENTS.

IA INPUT ARRAY TO BE CONVERTED TO  
QUANTIZED ARRAY  
NGL NUMBER OF GREY TONE LEVELS IN IA  
NQ NUMBER OF QUANTIZING LEVELS  
IASIZE SIZE OF ARRAY IA

KEQUAN14

KEQUAN15

KEQUAN16

KEQUAN17

KEQUAN18

C

SUBROUTINE KEQUAN(IA,NGL,NQ,IASIZE)

KEQUAN19

KEQUAN20

KEQUAN21

KEQUAN22

C

DIMENSION IA(1),KN(512)  
IF(NGL.GT.512) WRITE(6,10)  
10 FORMAT(5X,\*\*\*\*\*NUMBER OF GREY LEVELS TOO LARGE\*\*\*\*\*)

KEQUAN23

KEQUAN24

KEQUAN25

KEQUAN26

10

DO 1 I=1,NGL

KN(I)=0

COUNT EACH GREY LEVEL

KEQUAN27

KEQUAN28

KEQUAN29

1

DO 2 I=1,IASIZE

J=IA(I)

KN(J+1)=KN(J+1)+1

NP=IASIZE

J=1

MQ=NQ

GO THRU NQ LEVELS

KEQUAN30

KEQUAN31

KEQUAN32

KEQUAN33

C

DO 3 I=1,NQ

NL=NP

GET NEW LEVEL

KEQUAN34

KEQUAN35

KEQUAN36

KEQUAN37

C

NL=NL-MQ\*KN(J)

NP=NP-KN(J)

KN(J)=I-1

J=J+1

CHECK FOR LAST LEVEL

KEQUAN38

KEQUAN39

KEQUAN40

KEQUAN41

C

IF(J.GT.NGL) GO TO 6

INCREMENT AGAIN FOR LEVEL

KEQUAN42

KEQUAN43

KEQUAN44

C

IF(MQ\*KN(J).LE.NL\*2)GO TO 4

DECREASE NO. OF LEVELS LEFT

KEQUAN45

KEQUAN46

KEQUAN47

C

MQ=MQ-1

DO 5 I=J,NGL

KN(I)=NQ-1

GO TO 8

KEQUAN48

KEQUAN49

KEQUAN50

C

3

C

5

```
C
6 N=(NQ-I)/2
  IF(N.LT.1)GO TO 8
  DO 7 I=1,NG
  KN(I)=KN(I)+N
7
C
8 DO 9 I=1,ASIZE
  J=IA(I)
  IA(I)=KN(J+1)
9  RETURN
  END
```

SET LAST LEVEL

ASSIGN ELEMENTS TO A LEVEL AND RETURN

KEQUAN51  
KEQUAN52  
KEQUAN53  
KEQUAN54  
KEQUAN55  
KEQUAN56  
KEQUAN57  
KEQUAN58  
KEQUAN59  
KEQUAN60  
KEQUAN61

```

CRODSK2          READ DISC AND LIST GREY TONES
C
C              VERSION II WRITTEN BY RJ BOSLEY          JAN 1974
C
C      DESCRIPTION OF PROGRAM.
C      THIS SUBROUTINE READS DATA FROM DISC FILE NDSK AND PRINTS OUT
C      THE GREY TONES AND WRITES THEM OUT ONTO AN OUTPUT TAPE.
C
C      INTERNAL PARAMETERS.
C      K              BAND BEING PROCESSED
C      LINE          ERTS LINE BEING READ
C
C      ENTRY POINT.
C      CALL R0DSK2 (IMAGE,IRSTRT,IRSTOP,NHOR,NDSK,PRNT,TAPE,IFIL)
C
C      INPUT ARGUMENTS.
C      IMAGE          ARRAY TO READ DATA INTO
C      IRSTRT,IRSTOP  STARTING,STOPING LINES OF DATA
C
C      NHOR          NUMBER OF HORIZONTAL BLOCKS
C      NDSK          FILE CODE OF JISC WITH DATA
C      PPNT          TRUE FOR GREY-TONE LISTING
C      TAPE          TRUE FOR TAPE OUTPUT
C      IFIL          OUTPUT TAPE FILE CODE
C
C      SUBROUTINE R0DSK2 (IMAGE,IRSTRT,IRSTOP,NHOR,NDSK,PRNT,TAPE,IFIL)
C
C      DIMENSION IMAGE (41,41)
C      LOGICAL PRNT,TAPE
C      REWIND NDSK
C      K=NDSK-10
C
C      IF NEITHER TAPE NOR PRINT, RETURN
C
C      IF (PRNT) GO TO 2
C      IF (TAPE) GO TO 3
C      RETURN
C
C      WRITE HEADING FOR LIST
C      WRITE (6,1) K
C      FORMAT (1H1, 'LINE STRIP', 20X, 'BAND NUMBER IS', I2)
C      GO THRU EACH LINE OF DATA
C
C      DO 50 LINE=IRSTRT,IRSTOP
C      GO THRU EACH BLOCK ACROSS THE IMAGE
C
C      DO 50 J=1,NHOR
C      READ ONE LINE
C      READ (NDSK) (IMAGE (1,KOL),KOL=1,41)
C      WRITE IT OUT
C      IF (PRNT) WRITE (6,100) LINE,J,(IMAGE (1,KOL),KOL=1,41)
C      IF (TAPE) WRITE (IFIL) LINE,J,(IMAGE (1,KOL),KOL=1,41)
C
C      50 CONTINUE
C      100 FORMAT (1X,I5,I3,4I13)
C      RETURN
C      END
R0DSK201
R0DSK202
R0DSK203
R0DSK204
R0DSK205
R0DSK206
R0DSK207
R0DSK208
R0DSK209
R0DSK210
R0DSK211
R0DSK212
R0DSK213
R0DSK214
R0DSK215
R0DSK216
R0DSK217
R0DSK218
R0DSK219
R0DSK220
R0DSK221
R0DSK222
R0DSK223
R0DSK224
R0DSK225
R0DSK226
R0DSK227
R0DSK228
R0DSK229
R0DSK230
R0DSK231
R0DSK232
R0DSK233
R0DSK234
R0DSK235
R0DSK236
R0DSK237
R0DSK238
R0DSK239
R0DSK240
R0DSK241
R0DSK242
R0DSK243
R0DSK244
R0DSK245
R0DSK246
R0DSK247
R0DSK248
R0DSK249
R0DSK250

```

## IV.2-b Texture Analysis Program Listings

MAINLN

ERTS (see IV.2-a)

MAING

KEQUAN (see IV.2-a)

PITCHR (see IV.2-a)

FPLXIT

INDEX

IMOMTR

COR

IEQPQ1

RITOWT

CHAINLN	ERTS TEXTURE ANALYSIS	JUNE 1973	MAINLN01
C			MAINLN02
C	WRITTEN BY RJ BOSLEY FOR PROCESSING LAWRENCE DATA	SEPT 1972	MAINLN03
C	VERSION 1 BY RJ BOSLEY FOR PROCESSING ERTS DATA	NOV 1972	MAINLN04
C	VERSION 2 BY RJ BOSLEY FOR MERGE OPTION	JUNE 1973	MAINLN05
C	*****START OF DOCUMENTATION COMMENT CARDS*****		MAINLN06
*			MAINLN07
*			MAINLN08
*	PROGRAM IDENTIFICATION.		MAINLN09
*	PROGRAM TITLE - TEXTURE ANALYSIS		MAINLN10
*	WRITTEN BY - RJ BOSLEY		MAINLN11
*	DATE WRITTEN - JUNE 1973		MAINLN12
*	SITE NAME - CRINC, UNIV OF KANSAS		MAINLN13
*	SITE ADDRESS - LAWRENCE, KANSAS		MAINLN14
*	PHONE NUMBER - 1-913-864-4832		MAINLN15
*			MAINLN16
*			MAINLN17
*	DATE OF MODIFICATION.		MAINLN18
*			MAINLN19
*	JUNE 1973 -- MODIFIED FOR MERGE OPTION.		MAINLN20
*			MAINLN21
*	HARDWARE / SOFTWARE SUMMARY.		MAINLN22
*			MAINLN23
*	COMPUTER REQUIRED - HONEYWELL 635		MAINLN24
*	SYSTEM EXECUTIVE - GE OS III		MAINLN25
*	MEMORY REQUIRED - 42K		MAINLN26
*	PROGRAM LANGUAGE - FORTRAN IV AND GMAP		MAINLN27
*	PERIPHERALS - ONE SCRATCH DISK, TWO TAPE DRIVES		MAINLN28
*			MAINLN29
*			MAINLN30
*	PURPOSE.		MAINLN31
*	THIS TEXTURE ANALYSIS PACKAGE WAS WRITTEN IN ORDER TO		MAINLN32
*	PROCESS ERTS IMAGERY DATA USING PATTERN RECOGNITION TECHNIQUES.		MAINLN33
*			MAINLN34
*	METHOD.		MAINLN35
*	SEE 'LAND USE CLASSIFICATION', TECH. REPORT NO.2262-1, JANUARY 1973		MAINLN36
*			MAINLN37
*	INPUT.		MAINLN38
*			MAINLN39
*	PARAMETER CARDS.		MAINLN40
*	1. TITLE CARD. THIS CARD IS USED FOR TITLE INFORMATION		MAINLN41
*	AND IS LISTED ON THE OUTPUT LISTING.		MAINLN42
*	2. PARAMETERS ACCORDING TO THE FORTRAN NAMELIST FORMAT		MAINLN43
*	UNDER THE NAME PARAM, SEE BELOW.		MAINLN44
*			MAINLN45
*	OTHER INPUT.		MAINLN46
*			MAINLN47
*	ERTS IMAGE DATA TAPE ON INPUT FILE CODE 'ES'.		MAINLN48
*			MAINLN49
*	ABOPTS.		MAINLN50

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## SPECIFIED IN OUTPUT LISTING.

## OUTPUT.

## PRINTER OUTPUT.

1. PARAMETER VALUES.
2. PICTURE OF IMAGE BEING PROCESSED.
3. COORDINATES OF THE IMAGE.
4. VALUES OF THE EXTRACTED FEATURES.
5. LISTING OF LFX ARRAYS.

## CARD OUTPUT.

1. COORDINATES OF THE IMAGE BEING PROCESSED.
2. VALUES OF THE FEATURES EXTRACTED.
3. CARD COUNT.

## TAPE OUTPUT.

1. COORDINATES OF THE IMAGE BEING PROCESSED.
2. VALUES OF THE FEATURES EXTRACTED.

## RESTRICTIONS.

1. DATA SET IS STANDARD ERTS DATA TAPE FROM NASA.
2. THE NUMBER OF POINTS IN ONE LINE OF THE STRIP BEING PROCESSED MUST NOT EXCEED 192 POINTS.
3. THE SIZE OF EACH IMAGE MUST NOT EXCEED 4096 POINTS.
4. QUANTIZATION MUST BE 32 LEVELS OR LESS.
5. THE ERTS INPUT TAPE MUST HAVE A FILE CODE 'ES'.
6. IBAND MUST NOT EXCEED FOUR.

## SUBPROGRAMS REQUIRED.

```

MAINLN
  ERTS READ PROGRAM
  MAING
    KEQUAN
    PICTUR
    FPLXIT
      INDEX
    IMOMTR
      INDEX
      COR
      IEQP01
    RITOWT

```

## CARD SETUP FOR SAMPLE RUN.

```

$ IDENT 9999, ANYNAME
$ LIBRARY LB
$ OBJECT M-A-I-N-L-N
$ ... ERTS TEXTURE ANALYSIS PROGRAMS ...
$ JKEND
$ EXECUTE

```

```

MAINLN51
MAINLN52
MAINLN53
MAINLN54
MAINLN55
MAINLN56
MAINLN57
MAINLN58
MAINLN59
MAINLN60
MAINLN61
MAINLN62
MAINLN63
MAINLN64
MAINLN65
MAINLN66
MAINLN67
MAINLN68
MAINLN69
MAINLN70
MAINLN71
MAINLN72
MAINLN73
MAINLN74
MAINLN75
MAINLN76
MAINLN77
MAINLN78
MAINLN79
MAINLN80
MAINLN81
MAINLN82
MAINLN83
MAINLN84
MAINLN85
MAINLN86
MAINLN87
MAINLN88
MAINLN89
MAINLN90
MAINLN91
MAINLN92
MAINLN93
MAINLN94
MAINLN95
MAINLN96
MAINLN97
MAINLN98
MAINLN99
MAINLN01

```

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C		MAINLN01
C	READ TITLE AND WRITE IT OUT	MAINLN02
C		MAINLN03
	READ(5,6) (TITLE(I),I=1,14)	MAINLN04
	WRITE(6,21)	MAINLN05
	WRITE(6,7) (TITLE(I),I=1,14)	MAINLN06
21	FORMAT(////////40X,'ERTS TEXTURE ANALYSIS'////)	MAINLN07
6	FORMAT(13A6,A2)	MAINLN08
7	FORMAT(20X,13A6,A2////)	MAINLN09
C		MAINLN10
C	READ ERTS TFXTURE PARAMETERS	MAINLN11
C		MAINLN12
	READ(5,PAPAM)	MAINLN13
	M=NUMIM*NUMPPL	MAINLN14
	IF(M.LE.192) GO TO 2	MAINLN15
	WRITE(6,1)	MAINLN16
1	FORMAT(' NUMBER OF IMAGES TIMES THE NUMBER OF POINTS _V EACH LINE	MAINLN17
	1 MUST NOT EXCEED 192')	MAINLN18
	STOP	MAINLN19
C		MAINLN20
C	INITIALIZE THE ERTS READ PROGRAM AND WRITE OUT PARAMETERS	MAINLN21
C		MAINLN22
2	CALL EINIT(LENGTH)	MAINLN23
C		MAINLN24
	WRITE(6,11) LENGTH,NUMIM,NUMSTR,NBVERT,IBAND,NBSKIP	MAINLN25
	WRITE(6,101) PUNCH,NUMPPL,NUM_LIN,NRED,NSTART,NTIMES,NQUANT	MAINLN26
11	FORMAT(10X,'LENGTH OF ERTS LINE IS ',	MAINLN27
	1 I5,' POINTS'/10X,' NUMBER OF HORIZONTAL IMAGES,NUMIM,IS ',I2,/'	MAINLN28
	210X,' THIS STRIP IS NUMBER ',I2,/'10X,' NUMBER OF VERTICAL IMAGES COM	MAINLN29
	3TAINED IN STRIP IS ',I2,/'10X,' PROCESSING WILL BE ON BAND ',I2,/'	MAINLN30
	410X,' SKIPPED DOWN ',I3,' VERTICAL IMAGES BEFORE STARTING'////)	MAINLN31
101	FORMAT(10X,'PUNCH=',A3,' NUMPP_=',I4,' NUMLIN=',I4,' NRED=',I3,	MAINLN32
	1' NSTART=',I3,' NTIMES=',I3,' NQUANT=',I4)	MAINLN33
	IF(MERGE) WRITE(6,3)	MAINLN34
3	FORMAT(//10X,'THE FOUR _EX ARRAYS HAVE BEEN MERGED INTO ONE ARRAY'	MAINLN35
	1 //)	MAINLN36
	IF(.NOT.PICTURE) WRITE(6,4)	MAINLN37
4	FORMAT(' THE PICTURE OPTION IS OFF'////)	MAINLN38
C		MAINLN39
C	SKIP THE FIRST NBSKIP ROWS OF IMAGES	MAINLN40
C		MAINLN41
	NOSK=NUMLIN*NBSKIP	MAINLN42
	CALL ESKIP(NOSK)	MAINLN43
C		MAINLN44
C	GO DOWN THE STRIP	MAINLN45
C	IMAGE GOORDINATES (M1,N1) ARE TRANSFERRED IN COMMON	MAINLN46
C	M1 GIVES THE ROW COUNT GOING DOWN THE STRIP	MAINLN47
C		MAINLN48
	IBEGIN=((NUMSTR-1)*(192*4))+IBAND	MAINLN49
C		MAINLN50

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ORIGINAL PAGE IS POOR

```

C      MOVE THE IMAGE TO THE RIGHT BY EIGHT POINTS
C
      IBEGIN=IBFGIN+32
      JSTOP=IBEGIN+((NUMIM*NUMPPL)*4)-IBAND
      NB=NRSKIP+1
      DO 99 M1=NB,NBVERT
C
C      *****SECTION II--READ ERTS AND MOVE DATA INTO IMAGE*****
C      I, ILINE IS THE ARRAY INTO WHICH THE ERTS DATA IS READ--ILINE AND
C      IWORK USE THE SAME STORAGE SPACE
C      MDOWN GIVES THE ROW COUNT IN IMAGE FROM 1 DOWN TO NUMLIN
C
      DO 90 MDOWN=1,NUMLIN
C
      READ ERTS LINE BY LINE
C
      CALL EREAD(ILINE, LN)
C
      LN, RETURNED BY EREAD, GIVES THE LINE NUMBER, OR ERROR INDICATION
C
      IF (LN.EQ.0) GO TO 996
C
      MOVE EVERY POINT IN THE LINE THAT BELONGS TO IBAND, INTO IMAGE
      JSTOP GIVES THE STOPPING POINT IN ILINE FOR THE TRANSFER
      THE TRANSFER IS INCREMENTED BY 4, THE NUMBER OF BANDS
      LCOUNT GOES FROM 1 TO NUMPPL*NUMIM, GIVING THE LENGTH COUNT
C
      LCOUNT=0
      DO 80 IPOINT=IBEGIN,JSTOP,4
      LCOUNT=LCOUNT+1
      IMAGE(MDOWN,LCOUNT)=ILINE(IPOINT)
      CONTINUE
80    CONTINUE
C
      IMAGE IS FULL, START TEXTURE ANALYSIS
C
C      *****SECTION III--PROCESS IMAGE, BLOCK BY BLOCK*****
C      BLOCK IMAGE INTO NUMLIN X NUMPPL BLOCKS FOR PROCESSING
C
      KE=0
      DO 60 JBLOCK=1,NUMIM
      KS=KE+1
      KE=KS+NUMPPL-1
      KL=0
      DO 59 KLINE=1,NUMLIN
C
      KL GOES FROM 1 TO NBR OF PTS WHILE KCOL GIVES THE COLUMN COUNT
C
      MAINLN51
      MAINLN52
      MAINLN53
      MAINLN54
      MAINLN55
      MAINLN56
      MAINLN57
      MAINLN58
      MAINLN59
      MAINLN60
      MAINLN61
      MAINLN62
      MAINLN63
      MAINLN64
      MAINLN65
      MAINLN66
      MAINLN67
      MAINLN68
      MAINLN69
      MAINLN70
      MAINLN71
      MAINLN72
      MAINLN73
      MAINLN74
      MAINLN75
      MAINLN76
      MAINLN77
      MAINLN78
      MAINLN79
      MAINLN80
      MAINLN81
      MAINLN82
      MAINLN83
      MAINLN84
      MAINLN85
      MAINLN86
      MAINLN87
      MAINLN88
      MAINLN89
      MAINLN90
      MAINLN91
      MAINLN92
      MAINLN93
      MAINLN94
      MAINLN95
      MAINLN96
      MAINLN97
      MAINLN98
      MAINLN99
      MAINLN00

```

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	DO 58 KCOL=KS,KE	MAINLN0
	KL=KL+1	MAINLN0
	I_LINE(KL)=IMAGE(K_LINE,KCOL)	MAINLN0
58	CONTINUE	MAINLN0
59	CONTINUE	MAINLN0
	N1=(N11-1)+JBLOCK	MAINLN0
C		MAINLN0
C	USE ILINE AS A DUMMY ARRAY TO SEND WORK TO MAING	MAINLN0
C		MAINLN0
C	CALL MAING(ILINE,MERR,MERGE,PICTUR,IF)	MAINLN1
C		MAINLN1
C	CHECK FOR ERROR CONDITION IN MAING	MAINLN1
C		MAINLN1
	IF(MERR.EQ.1) GO TO 992	MAINLN1
60	CONTINUE	MAINLN1
99	CONTINUE	MAINLN1
C		MAINLN1
C	*****SECTION IV--END OF STRIP, FINISH UP*****	MAINLN1
C		MAINLN1
C	PUT AN EOF MARK ON OUTPUT FILE IF AND WRITE ANOTHER RECORD	MAINLN2
C		MAINLN2
	ENDFILE IF	MAINLN2
	WRITE(IF) (ILINE(K),K=1,10)	MAINLN2
	STOP	MAINLN2
C		MAINLN2
C	ERROR DETECTED---WRITE FILE MARK AND ANOTHER RECORD	MAINLN2
C		MAINLN2
992	WRITE(6,993) M1,N1	MAINLN2
993	FORMAT(10X,'ERROR IN SUBROUTINE MAING, LAST IMAGE WAS (',I2,',',	MAINLN2
	I2,',')')	MAINLN3
	ENDFILE IF	MAINLN3
	WRITE(IF) (ILINE(K),K=1,10)	MAINLN3
	STOP	MAINLN3
C		MAINLN3
C	ERROR DETECTED---WRITE FILE MARK AND ANOTHER RECORD	MAINLN3
C		MAINLN3
996	WRITE(6,997) M1	MAINLN3
997	FORMAT(10X,'UNEXPECTED EOF ON ERTS, LAST ROW COMPLETED WAS ',I3)	MAINLN3
	ENDFILE IF	MAINLN3
	WRITE(IF) (ILINE(K),K=1,10)	MAINLN4
	STOP	MAINLN4
	END	MAINLN4

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```

CMAING          ERTS M-A-I-N-G          JJNE 1973          MAING001
C              WRITTEN BY   RMH          DATE OCT 1971          MAING002
C              VERSION 1 BY RJ BOSLEY FOR LAWRENCE DATA          SEPT 1973          MAING003
C              VERSION 2 BY RJ BOSLEY FOR ERTS DATA PROCESSING          NOV 1973          MAING004
C              VERSION 3 BY RJ BOSLEY FOR MERGE OPTION          JUNE 1973          MAING006
C              MAING007
C              MAING008
C              DESCRIPTION OF PPROGRAM.          MAING009
C              THIS SUPROUTINE PREPARES THE IMAGE IN ARRAY IWORK FOR PITCHR          MAING010
C              AND THEN PROCESSES IT, ACCORDING TO THE MERGE OPTION, CALLING          MAING011
C              SUBROUTINES FOR THE LEX ARRAYS, CALCULATING THE TEXTURE FEATURES          MAING012
C              AND THEN WRITTING OUT THE RESULTS.          MAING013
C              MAING014
C              ENTRY POINT.          MAING015
C              CALL MAING(IWORK,MERR,MERGE,PICTUR,IF)          MAING016
C              ARGUMENTS.          MAING017
C              IWORK          THE NUMLIN*NUMPP_ IMAGE ARRAY          MAING018
C              MERR          ERROR FLAG FOR LEX ARRAY SIZE          MAING019
C              MERGE          OPTION TO MERGE THE FOUR LEX ARRAYS          MAING020
C              PICTUR          OPTION TO PRINT PICTURE OF THE IMAGE          MAING021
C              IF          FILE CODE FOR OUTPUT TAPE IN RITOWT          MAING022
C              MAING023
C              INTERNAL PARAMETERS.          MAING024
C              NUMLIN          THE NUMBER OF LINES IN THE IMAGE          MAING025
C              NUMPP_          THE NUMBER OF POINTS PER IMAGE LINE          MAING026
C              NUMIM          MAXIMUM ALLOWABLE SIZE OF THE LEX          MAING027
C              ARRAYS, NUMPP_*NUMLIN          MAING028
C              NQ          NUMBER OF QUANTIZING LEVELS FOR KFQUANMAING029
C              IMAX          MAXIMUM GREY TONE LEVEL IN IWORK          MAING030
C              IMIN          MINIMUM GREY TONE LEVEL IN IWORK          MAING031
C              IS          STARTING PT OF A ROW OF IWORK          MAING032
C              IE          ENDING POINT OF THE ROW IN IWORK          MAING033
C              NGL          NUMBER OF GREY TONE LEVELS IN IWORK          MAING034
C              IASIZE          SIZE OF IWORK ARRAY          MAING035
C              HORZ          HORIZONTAL SCALE FACTOR FOR PITCHR          MAING036
C              VERT          VERTICAL SCALE FACTOR FOR PITCHR          MAING037
C              LEAST1          ONE GREY TONE LEVEL BELOW IMIN          MAING038
C              NOBL          NUMBER OF GREY TONE LEVELS IN IWORK          MAING039
C              NBURL          THE NUMBER OF LEVELS IN THE TRIANGULA          MAING040
C              LEX ARRAY          MAING041
C              L1...L6          ADDRESS INDEXS FOR THE LEX ARRAYS          MAING042
C              MAING043
C              MAING044
C              MAING045
C              MAING046
C              MAING047
C              MAING048
C              MAING049
C              MAING050
C              SUBROUTINE MAING(IWORK,MERR,MERGE,PICTUR,IF)          MAING051
C              LOGICAL MERGE,PICTUR          MAING052
C              DIMENSION IWORK(4096)          MAING053

```

REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR

```

DIMENSION G(64),IQ(64)
COMMON /Q/ NQUANT
COMMON M1,N1,F(15),IMAX,IMIN,NUMPPL,NUMLIN,NBU9L,IR1,IR2,IR3,IR4,
1 DUMMY(29),LEAST1,NPED,N_LAYER,NSTART,NTIMES,NJ,PNC4
COMMON /E/ENTROP(4),DIFENT(4),DLFAVE(4),JIFVAR(4),SUMENT(4),
1SUMAVE(4),SUMVAR(4)
COMMON /CORREL/CORINF(4),CORMUT(4),CORMAX(4)
COMMON IMAGE(64,192)
NDIM=NUMPPL*NUMLIN
NQ=32
REWIND SCRATCH FILE, COPY IMAGE IN A LINE BY LINE FASHION ONTO
THE SCRATCH FILE, AND DETERMINE THE MINIMUM AND MAXIMUM GREY TONE
REWIND 2
FIRST, QUANTIZE THE ARRAY TO NQ LEVELS FOR PICTURE AND EFFICIFNCY
IMAX=-10000
IMIN=10000
DO 13 J=1,NUMLIN
IS=NUMPPL*(J-1)+1
IE=NUMPPL*J
DO 12 K=IS,IE
IF (IMIN.GT.IWORK(K)) IMIN=IWORK(K)
IF (IMAX.LT.IWORK(K)) IMAX=IWORK(K)
12 CONTINUE
13 CONTINUE
NGL=IMAX+1
IASIZE=NUMPPL*NUMLIN
CALL KEQUAN(IWORK,NGL,NQ,IASIZE)
COPY IMAGE ON SCRATCH FILE AFTER QUANTIZATION
DO 20 I=1,NUMLIN
IS=NUMPPL*(I-1)+1
IE=NUMPPL*I
WRITE(2) (IWORK(K),K=IS,IE)
CONTINUE
THE MAXIMUM AFTER KEQUAN QUANTIZES TO NQ LEVELS IS NQ-1,MINIMUM=0
IMAX=NQ-1
IMIN=0
TEST FOR PICTURE OPTION
IF (.NOT.P.PCTUR) GO TO 16
TRANSPOSE IWORK FOR PITCHR

```

MAING051  
 MAING052  
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 MAING055  
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 MAING095  
 MAING096  
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 MAING098  
 MAING099  
 MAING100

REPRODUCIBILITY OF THE  
 ORIGINAL PAGE IS POOR

		MAING101
		MAING102
		MAING103
		MAING104
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		MAING141
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		MAING144
		MAING145
		MAING146
		MAING147
		MAING148
		MAING149
		MAING150

```

C
  L=0
  DO 14 I=1,NUMLIN
  L=L+1
  IS=NUMPPL*(I-1)+1
  IE=NUMPPL*I
  K=0
  DO 15 J=IS,IE
  K=K+1
  IMAGE(L,K)=IWORK(J)
15 CONTINUE
CONTINUE
14
C
C
C
  PRINT OUT PICTURE OF THE IMAGE
C
  HORZ=(64.0*0.90)/FLOAT(NUMPPL)
  VERT=(64.0*0.75)/FLOAT(NUMLIN)
  CALL PITCHR(IMAGE,NUMLIN,NUMPPL,0,1,0,IMAX,,0,,,HORZ,VERT, )
  LEAST1=IMIN-1
16
C
C
C
  NOBL IS THE NUMBER OF BRIGHTNESS LEVELS
C
  NOBL=IMAX-LEAST1
C
  NBUBL IS THE NUMBER OF LEVELS IN THE TRIANGULAR LEX ARRAY
C
  NBUBL=NOBL*(NOBL+1)/2
C
C
C
  SET UP THE INDEXS FOR THE LEX ARRAYS
C
  L1=1
  L2=L1+NUMPPL*2
  L3=L2+NBUBL
  L4=L3+NBUBL
  L5=L4+NBUBL
  L6=L5+NBUBL-1
C
C
C
  CHECK THE SIZE
C
  IF(L6.GT.NDIM) GO TO 78
  DO 4 NN=NSTART,NTIMES
  REWIND 2
  NLAYER=NN-1
C
C
C
  GET THE LEX ARRAYS
C
  CALL FPLXIT(IWORK(L1),IWORK(L2),IWORK(L3),IWORK(L4),IWORK(L5),
1  NUMPPL,MERGE)
C
C
  CALCULATE THE TEXTURE FEATURES

```

```
C      CALL IMONTR(IWORK(L2),IWORK(L3),IWORK(L4),IWORK(L5),G,I7,MERGE)      MAING15
C      OUTPUT THE TEXTURE DATA      MAING15
C      CALL RITOWT(IWORK(L2),IWORK(L3),IWORK(L4),IWORK(L5),G,I7,MERGE,IF,      MAING15
1 PICTUP)      MAING15
4 CONTINUE      MAING15
C      SET ERROR INDICATOR TO NO ERRORS      MAING16
C      MERR=0      MAING16
      RETURN      MAING16
78 WRITE(6,104) NDIM,L6      MAING16
104 FORMAT(6H NDIM=,I5,16H NDIM MUST BE = ,I7)      MAING16
      MERR=1      MAING16
      RETURN      MAING16
      END      MAING16
```

```

CFPLXIT                                F-P-L-X-I-T                                JUNE 1973                                FPLXIT01
C                                        WRITTEN BY RMH                                SEPT 1971                                FPLXIT02
C                                        VERSION 1 BY RJ ROSLEY FOR MERGE          JUNE 1973                                FPLXIT03
C                                                                              FPLXIT04
C                                                                              FPLXIT05
C                                                                              FPLXIT06
C                                                                              FPLXIT07
C                                                                              FPLXIT08
C                                                                              FPLXIT09
C                                                                              FPLXIT10
C                                                                              FPLXIT11
C                                                                              FPLXIT12
C                                                                              FPLXIT13
C                                                                              FPLXIT14
C                                                                              FPLXIT15
C                                                                              FPLXIT16
C                                                                              FPLXIT17
C                                                                              FPLXIT18
C                                                                              FPLXIT19
C                                                                              FPLXIT20
C                                                                              FPLXIT21
C                                                                              FPLXIT22
C                                                                              FPLXIT23
C                                                                              FPLXIT24
C                                                                              FPLXIT25
C                                                                              FPLXIT26
C                                                                              FPLXIT27
C                                                                              FPLXIT28
C                                                                              FPLXIT29
C                                                                              FPLXIT30
C                                                                              FPLXIT31
C                                                                              FPLXIT32
C                                                                              FPLXIT33
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C                                                                              FPLXIT38
C                                                                              FPLXIT39
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C                                                                              FPLXIT41
C                                                                              FPLXIT42
C                                                                              FPLXIT43
C                                                                              FPLXIT44
C                                                                              FPLXIT45
C                                                                              FPLXIT46
C                                                                              FPLXIT47
C                                                                              FPLXIT48
C                                                                              FPLXIT49
C                                                                              FPLXIT50

C
C      WRITTEN BY RMH
C      VERSION 1 BY RJ ROSLEY FOR MERGE
C
C      DESCRIPTION OF PROGRAM.
C      THIS SUBROUTINE COMPUTES FOUR NEAREST NEIGHBOR GREY TONE MA-
C      TRICES, LEX1, LEX2, LEX3, AND LEX4 FOR ANGLES OF 90-DEGREES,
C      0-DEGREES, 135-DEGREES, AND 45-DEGREES RESPECTIVELY.
C      INCLUDED IN THIS SUBROUTINE IS AN OPTION TO MERGE THE FOUR LEX
C      ARRAYS INTO ONE, LEX1
C
C      ENTRY POINT.
C      CALL FPLXIT(IDATA,LEX1,LEX2,LEX3,LEX4,NUMPPL,MERGE)
C
C      ARGUMENTS.
C      IDATA          WORKING ARRAY FOR TWO LINES OF IMAGE
C      LEX1-LEX4      ADDRESS INDEXES FOR LEX ARRAYS
C      NUMPPL         NUMBER OF PTS PER IMAGE LINE
C      MERGE          OPTION TO MERGE THE FOUR LEX ARRAYS
C                   INTO ONE ARRAY, LEX1
C
C      INTERNAL PARAMETERS.
C      NBU3L         NUMBER OF LEVELS IN THE TRIANGULAR
C                   LEX ARRAYS
C      IST          POINTER TO FIRST LINE
C      NND          POINTER TO SECOND LINE
C      NRED         BASE FOR IMAGE REDUCTION
C      NLayer       THE POWER TO WHICH NRED IS RAISED
C      MM          AMOUNT OF REDUCTION OF THE IMAGE
C      FILE 2      SCRATCH FILE CONTAINING THE IMAGE
C      I,J,L,K     GREY TONE VALUES OF NEIGHBORING
C                   RESOLUTION CELLS, ONE TO EACH ANGLE
C      INDEX(I,J)  FUNCTION USED TO RETURN A SINGLE
C                   SUBSCRIPT FOR THE LEX ARRAY
C                   INDICATING WHERE ELEMENT (I,J) CAN
C                   BE FOUND
C
C      SUBROUTINE FPLXIT(IDATA,LEX1,LEX2,LEX3,LEX4,NUMPPL,MERGE)
C      DIMENSION IDATA(NUMPPL,2),LEX1(1),LEX2(1),LEX3(1),LEX4(1)
C      COMMON M1,N1,TYPE,F(14)
C      COMMON I1,I2,NUMPPL,NUMLIN,NBU3L,IR1,IR2,IR3,IR4,DUMMY(29)
C      COMMON LEAST1,NRED,NLayer,NSTART,NTIMES
C      LOGICAL MERGE
C
C      INITIALIZE LEX1, LEX2, LEX3, AND LEX4 ARRAYS TO ZERO
C      DO 10 I = 1,NBU3L

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      LEX1(I) = 0
      LEX2(I) = 0
      LFX3(I) = 0
10    LEX4(I) = 0
C
      IR1 = 0
      IR2 = 0
      IR3 = 0
      IR4 = 0
C
      IST POINTS TO FIRST LINE. NND POINTS TO SECOND LINE
C
      IST = 1
      NND = ?
C
      N LAYER INDICATES BY HOW MUCH THE IMAGE WILL BE REDUCED.
      NRED IS THE FACTOR BY WHICH THE IMAGE WILL BE REDUCED. (IT IS THE
      BASE WHICH IS RAISED TO THE POWER N LAYER.) THEN, BY DEFINING THE
      QUANTITY MM, WHERE MM = NRED**N LAYER, WE HAVE A SINGLE FACTOR
      THAT DETERMINES THE REDUCTION BASE AND THE AMOUNT OF THE REDUC-
      ION. IF, FOR EXAMPLE, NRED = 2, AND N LAYER RANGES FROM 0 TO 3 --
      THIS RANGE IS DETERMINED BY THE PARAMETER N TIMES (SEE *MAIN*),
      THE RESULTANT PROCESSING WILL YIELD FOUR IMAGES THAT WILL BE SUC-
      CESSIVELY REDUCED BY 1, 1/2, 1/4, AND 1/8 RESPECTIVELY.
C
      MM = NRED**N LAYER
      NUMPPL2 = NUMPPL/MM
      DO 111 KK1=1,MM
      DO 111 KK2=1,MM
C
      GET THE FIRST LINE OF DATA FROM DISC FILE 02
C
      DO 8 LL=1, KK1
      8  READ(2) (IDATA(L,IST),L=1,NUMPPL)
         N = 0
      DO 29 J=KK2,NUMPPL,MM
         N = N + 1
      29 IDATA(N,IST) = IDATA(J,IST)
C
      MMM=MM+KK1
      DO 1 LCNT = MMM,NUMLIN,MM
C
      GET THE SECOND LINE OF DATA. AFTER EACH ITERATION, THE OLD SEC-
      OND LINE BECOMES THE NEW FIRST LINE.
C
      DO 18 LL=1,MM
      18 READ(2) (IDATA(L,NND),L=1,NUMPPL)
         N = 0
      DO 19 J=KK2,NUMPPL,MM
         N = N + 1

```

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FPLXIT51
FPLXIT52
FPLXIT53
FPLXIT54
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FPLXIT72
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FPLXIT90
FPLXIT91
FPLXIT92
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FPLXIT96
FPLXIT97
FPLXIT98
FPLXIT99
FPLXIT00

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	IL = INDEX(I,L)	FPLXIT01
C	COUNT VERTICALLY ADJACENT (90-DEGREE) NEAREST NEIGHBORS.	FPLYIT02
C		FPLXIT03
	LEX1(IL) = LEX1(IL) + 1	FPLXIT04
	IR1 = IR1 + 1	FPLXIT05
C		FPLXIT06
	IJ = INDEX(I,J)	FPLXIT07
C	COUNT HORIZONTALLY ADJACENT (0-DEGREE) NEAREST NEIGHBORS.	FPLXIT08
C		FPLXIT09
	LEX2(IJ) = LEX2(IJ) + 1	FPLXIT10
	IR2 = IR2 + 1	FPLXIT11
C		FPLXIT12
	IK = INDEX(I,K)	FPLXIT13
C	COUNT 'LEFT DIAGONALLY' ADJACENT (135-DEGREE) NEAREST NEIGHBORS.	FPLXIT14
C		FPLXIT15
	LEX3(IK) = LEX3(IK) + 1	FPLXIT16
	IR3 = IR3 + 1	FPLXIT17
C		FPLXIT18
	IM = INDEX(I,M)	FPLXIT19
C	COUNT 'RIGHT DIAGONALLY' (45-DEGREE) ADJACENT NEAREST NEIGHBORS.	FPLXIT20
C		FPLXIT21
	LEX4(IM) = LEX4(IM) + 1	FPLXIT22
	IR4 = IR4 + 1	FPLXIT23
C		FPLXIT24
2	CONTINUE	FPLXIT25
C		FPLXIT26
	MAKE COUNT FOR LAST COLUMN.	FPLXIT27
C		FPLXIT28
	I = J	FPLXIT29
	M = L	FPLXIT30
	L = K	FPLXIT31
C		FPLXIT32
	IL = INDEX(I,L)	FPLXIT33
C	COUNT VERTICALLY ADJACENT NEAREST NEIGHBORS FOR LAST COLUMN.	FPLXIT34
C		FPLXIT35
	LEX1(IL) = LEX1(IL) + 1	FPLXIT36
	IR1 = IR1 + 1	FPLXIT37
C		FPLXIT38
	IM = INDEX(I,M)	FPLXIT39
C	COUNT 'RIGHT DIAGONALLY' ADJACENT NEAREST NEIGHBORS FOR THE LAST	FPLXIT40
C	COLUMN.	FPLXIT41
	LEX4(IM) = LEX4(IM) + 1	FPLXIT42
	IR4 = IR4 + 1	FPLXIT43
C		FPLXIT44
		FPLXIT45
C		FPLXIT46
		FPLXIT47
C		FPLXIT48
		FPLXIT49
		FPLXIT50

```

C
C
C INTERCHANGE THE LINE POINTERS.
C
C MN = IST
C IST = NNO
C NNO = MN
C
C 1 CONTINUE
C
C MAKE COUNT FOR LAST ROW.
C
C I=IDATA(1,IST) - LEAST1
C J=IDATA(2,IST) - LEAST1
C
C IJ = INDEX(I,J)
C
C COUNT HORIZONTALLY ADJACENT NEAREST NEIGHBORS FOR FIRST TWO COL-
C UMS OF LAST ROW.
C
C LEX2(IJ) = LEX2(IJ) + 1
C IR2 = IR2 + 1
C
C COMPLETE COUNT FOR LAST ROW.
C
C DO 12 N = 3,NUMPL2
C I = J
C J = IDATA(N,IST) - LEAST1
C
C IJ = INDEX(I,J)
C
C COUNT HORIZONTALLY ADJACENT NEAREST NEIGHBORS FOR REMAINDER OF
C LAST ROW.
C
C LEX2(IJ) = LEX2(IJ) + 1
C IR2 = IR2 + 1
C 12 CONTINUE
C REWIND 02
C 111 CONTINUE
C
C NOW DOUBLE THE DIAGONAL TO MAKE EVERYTHING COME OUT RIGHT
C
C NOBL=I1-I2+1
C DO 100 I=1,NOBL
C II=INDEX(I,I)
C LEX1(II)=2*LEX1(II)
C LEX2(II)=2*LEX2(II)
C LEX3(II)=LEX3(II)*2
C LEX4(II)=LEX4(II)*2
C 100 CONTINUE

```

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FPLXIT51
FPLXIT52
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FPLXIT98
FPLXIT99
FPLXIT00

```

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ORIGINAL PAGE IS POOR

```

C      IF (.NOT.MERGE) RETURN
C      IF MFRGE IS TRUE, SUM ALL APRAYS INTO LEX1
C
DO 112 I=1,NBU9L
LEX1(I)=LEX1(I)+LEX2(I)+LEX3(I)+LEX4(I)
112 CONTINUE
      RETURN
      END

```

FPLXITO  
FPLXITO  
FPLXITO  
FPLXITO  
FPLXITO  
FPLXITO  
FPLXITO  
FPLXITO

769 WORDS OF MEMORY USED BY THIS COMPILATION

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

CINDEX

I-N-D-E-X

SEPT 1971

C  
C  
C  
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C  
C

WRITTEN BY RMH

GIVEN THE ROW AND COLUMN SUBSCRIPTS I AND L, INDEX RETURNS THE SINGLE SUBSCRIPT FOR THE LEX ARRAY INDICATING WHERE ELEMENT (I,L) CAN BE FOUND.

1				
2	3			
4	5	6		
7	8	9	10	

```

FUNCTION INDEX(I,L)
  INDX1(I,L) = (I-1)*I/2 + L
  IF(I.GT.L) GO TO 1
  INDEX = INDX1(L,I)
  RETURN
1 INDEX = INDX1(I,L)
  RETURN
END

```

INDEX001  
 INCFX002  
 INDEX003  
 INDFX004  
 INDEX005  
 INDEX006  
 INDEX007  
 INDEX008  
 INCFX009  
 INDEX010  
 INDEX011  
 INDEX012  
 INDEX013  
 INDEX014  
 INDEX015  
 INDEX016  
 INDEX017  
 INDEX018  
 INDEX019  
 INDEX020  
 INDEX021  
 INDEX022

REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR

CZIMOMTR

I-M-O-M-T-R

JUNE 1973

IMOMTR01

IMOMTR02

SEPT 1971

IMOMTR03

JUNE 1972

IMOMTR04

JUNE 1973

IMOMTR05

IMOMTR06

IMOMTR07

IMOMTR08

IMOMTR09

IMOMTR10

IMOMTR11

IMOMTR12

IMOMTR13

IMOMTR14

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IMOMTR23

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IMOMTR31

IMOMTR32

IMOMTR33

IMOMTR34

IMOMTR35

IMOMTR36

IMOMTR37

IMOMTR38

IMOMTR39

IMOMTR40

IMOMTR41

IMOMTR42

IMOMTR43

IMOMTR44

IMOMTR45

IMOMTR46

IMOMTR47

IMOMTR48

IMOMTR49

IMOMTR50

WRITTEN BY RMH  
 VERSION 1 BY SAM SHANMUGAM FOR LAWRENCE DATA  
 VERSION 2 BY RJ BOSLEY FOR MERGE OPTION

## DESCRIPTION OF PROGRAM.

THIS PROGRAM CALCULATES THE MOMENT TEXTURE STATISTICS (AS  
 DEFINED BELOW UNDER TEXTURAL FEATURES) FROM THE LEX ARRAYS,  
 ACCORDING TO THE MERGE OPTION.

## ENTRY POINT.

CALL IMOMTR(LEX1,LEX2,LEX3,LEX4,F,IQ,MERGE)

## ARGUMENTS.

LEX1-LEX4

F

IQ

MERGE

ADDRESS INDEXES FOR LEX ARRAYS  
 CUMULATIVE DISTRIBUTION FUNCTION  
 QUANTIZED OUTPUT ARRAY OF IEQPQ1  
 OPTION TO MERGE THE FOUR LEX ARRAYS  
 INTO ONE ARRAY

## INTERNAL PARAMETERS.

ANGMOM...CORMAX

NQUANT

NQUAN2

IMAX

IMIN

IR1-IR4

R1-R4

QD

P(XY)

NOBL

NADD1-NADD4

TEXTURAL FEATURES--SEE BELOW  
 NUMBER OF QUANTIZING LEVELS FOR IEQPQ1  
 TWICE NQUANT  
 MAXIMUM NUMBER OF GREY TONE LEVELS  
 MINIMUM NUMBER OF GREY TONE LEVELS  
 THE NUMBER OF RESOLUTION CELL PAIRS  
 INVERSE OF IR1-IR4  
 COUNTED IN EACH LEX ARRAY  
 SCRATCH ARRAY USED BY SUBROUTINE COR  
 ARRAY OF JOINT PROBABILITIES  
 NUMBER OF GREY TONE LEVELS  
 SUM OF ELEMENTS OF THE LEX ARRAY

## TEXTURAL FEATURES.

ANGMOM= SUM SUM P(I,J)\*P(I,J)  
 I J

AMEAN= SUM SUM I\*P(I,J)  
 I J

AMEAN= SUM SUM J\*P(I,J)  
 I J

SGMASQ= SUM SUM ((I-AMEAN)\*\*2)\*P(I,J)

REPRODUCIBILITY OF THE  
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C  
C  
C  
C  
C

IF MERGE IS .TRUE., THE FOUR LEX ARRAYS HAVE BEEN MERGED INTO LEX1.

SUBROUTINE IMOMTR(LEX1,LEX2,LEX3,LEX4,F,IQ,MERGE)

REAL IVDMMOM  
LOGICAL MERGE  
DIMENSION LEX1(1),LEX2(1),LEX3(1),LEX4(1)  
DIMENSION F(64),IQ(64)  
DIMENSION DIF1(64),DIF2(64),DIF3(64),DIF4(64)  
DIMENSION SUM1(128),SUM2(128),SUM3(128),SUM4(128)  
DIMENSION P(3600),Q(300)  
EQUIVALENCE(P(1),Q(1))  
COMMON /Q/ NQUANT  
COMMON M1,N1,TYPE,G(14)  
COMMON IMAX,IMIN,NUMPPL,NUMLIN,NBUBL,IR1,IR2,IR3,IR4,ANGMOM(4),  
1 AMEAN(4),SGMASQ(4),SGMAXY(4),DIFMOM(4),RATIO(4),IVDMOM(4),TMEAN  
COMMON /FC/ENTROP(4),DIFENT(4),DIFAVE(4),DIFVAR(4),SUMENT(4),  
1SUMAVE(4),SUMVAR(4)  
COMMON /CORREL/CORINF(4),CORMUT(4),CORMAX(4)

C  
C  
C

INITIALIZE ARRAYS TO ZERO

DO 1 I=1,4  
IVDMOM(I)=0  
ANGMOM(I)=0  
AMEAN(I)=0  
SGMASQ(I)=0  
SGMAXY(I)=0  
ENTROP(I)=0.0  
DIFENT(I)=0.0  
DIFAVE(I)=0.0  
DIFVAR(I)=0.0  
SUMENT(I)=0.0  
SUMAVE(I)=0.0  
SUMVAR(I)=0.0  
1 RATIO(I)=0  
DO 86 K=1,NQUANT  
DIF1(K)=0.0  
DIF2(K)=0.0  
DIF3(K)=0.0  
86 DIF4(K)=0.0  
NQUANT2=2\*NQUANT  
DO 87 KS=1,NQUANT2  
SUM1(KS)=0.0  
SUM2(KS)=0.0  
SUM3(KS)=0.0  
87 SUM4(KS)=0.0

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```

GET THE NUMBER OF BRIGHTNESS LEVELS,NOBL
IF THE LEX ARRAY WERE SQUARE AND NOT COMPACTED, IT WOULD BE
DIMENSIONED NOBL BY NOBL

NOBL=IMAX-IMIN+1

NOW DETERMINE THE TOTAL NUMBER OF RESOLUTION CELL PAIRS
COUNTED IN EACH OF THE LEX ARRAYS

IP1=0
IP2=0
IP3=0
IP4=0
IF(.NOT.MERGE) GO TO 40
DO 42 I=1,NOBL
DO 42 J=1,NOBL
IJ=INDEX(I,J)
IR1=IR1+LEX1(IJ)
CONTINUE
R1=1./FLOAT(IR1)
GO TO 41
DO 5 I=1,NOBL
DO 5 J=1,NOBL
IJ=INDEX(I,J)
IR1=IR1+LEX1(IJ)
IR2=IR2+LEX2(IJ)
IR3=IR3+LEX3(IJ)
5 IR4=IR4+LEX4(IJ)

GET R1,R2,R3,R4 TO SAVE DIVISIONS

R1=1./FLOAT(IP1)
R2=1./FLOAT(IP2)
R3=1./FLOAT(IP3)
R4=1./FLOAT(IP4)

FIND THE CORRELATION MEASURES
PUT THE LEX ARRAYS IN P MATRIX AND CALL CORRELATION ROUTINE

DO LEX2 ARRAY

JJ=0
DO 201 I=1,NOBL
DO 201 J=1,NOBL
IJ=INDEX(I,J)
JJ=JJ+1
201 P(JJ)=FLOAT(LEX2(IJ))*R1
CALL COR(P,NOBL,1,QD,COR1,COR2,COR3)
CORINF(1)=COR1

```

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	CORMUT(1)=COR2	IMOMTR01
	CORMAX(1)=COR3	IMOMTR02
	JJ=0	IMOMTR03
C		IMOMTR04
C	DO LEX4 ARRAY	IMOMTR05
C		IMOMTR06
	DO 211 I=1,NOBL	IMOMTR07
	DO 211 J=1,NOBL	IMOMTR08
	IJ=INDEX(I,J)	IMOMTR09
	JJ=JJ+1	IMOMTR10
211	P(JJ)=FLOAT(LEX4(IJ))*R2	IMOMTR11
	CALL COR(P,NOBL,1,QD,COR1,COR2,COR3)	IMOMTR12
	CORINF(2)=COR1	IMOMTR13
	CORMUT(2)=COR2	IMOMTR14
	CORMAX(2)=COR3	IMOMTR15
41	JJ=0	IMOMTR16
C		IMOMTR17
C	DO LEX1 ARRAY	IMOMTR18
C		IMOMTR19
	DO 221 I=1,NOBL	IMOMTR20
	DO 221 J=1,NOBL	IMOMTR21
	IJ=INDEX(I,J)	IMOMTR22
	JJ=JJ+1	IMOMTR23
221	P(JJ)=FLOAT(LEX1(IJ))*R1	IMOMTR24
	CALL COR(P,NOBL,1,QD,COR1,COR2,COR3)	IMOMTR25
	CORINF(3)=COR1	IMOMTR26
	CORMUT(3)=COR2	IMOMTR27
	CORMAX(3)=COR3	IMOMTR28
	IF(MERGE) GO TO 43	IMOMTR29
	JJ=0	IMOMTR30
C		IMOMTR31
C	DO LEX3 ARRAY	IMOMTR32
C		IMOMTR33
	DO 231 I=1,NOBL	IMOMTR34
	DO 231 J=1,NOBL	IMOMTR35
	IJ=INDEX(I,J)	IMOMTR36
	JJ=JJ+1	IMOMTR37
231	P(JJ)=FLOAT(LEX3(IJ))*R4	IMOMTR38
	CALL COR(P,NOBL,1,QD,COR1,COR2,COR3)	IMOMTR39
	CORINF(4)=COR1	IMOMTR40
	CORMUT(4)=COR2	IMOMTR41
	CORMAX(4)=COR3	IMOMTR42
C		IMOMTR43
C	GET THE PROBABILITY FUNCITON IN F	IMOMTR44
C		IMOMTR45
43	DO 379 I=1,64	IMOMTR46
	Q(I)=0	IMOMTR47
379	F(I)=0	IMOMTR48
C		IMOMTR49
C	IF MERGE. GO TO SECTION AI TO MAKE THE COMPUTATIONS	IMOMTR50

REPRODUCIBILITY OF THE  
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C	IF (MERGE) GO TO 911	IMOMTR51
	DO 6 I=1,NOBL	IMOMTR52
	IA=0	IMOMTR53
	DO 7 J=1,NOBL	IMOMTR54
	IJ=INDEX(I,J)	IMOMTR55
	7 IA=IA+LEX1(IJ)+LEX2(IJ)+LEX3(IJ)+LEX4(IJ)	IMOMTR56
	6 F(I)=FLOAT(IA)/FLOAT(IR1+IR2+LR3+IR4)	IMOMTR57
C		IMOMTR58
C	FIRST COMPUTE THE TRUE MEAN	IMOMTR59
C		IMOMTR60
	TMEAN=0	IMOMTR61
	DO 10 I=1,NORL	IMOMTR62
	10 TMEAN=TMEAN+F(I)*FLOAT(I)	IMOMTR63
	TMEAN=TMEAN/FLOAT(IMIN-1)	IMOMTR64
C		IMOMTR65
C	GET CUMULATIVE DISTRIBUTION FUNCTION IN F	IMOMTR66
C		IMOMTR67
	DO 8 I=2,NOBL	IMOMTR68
	8 F(I)=F(I)+F(I-1)	IMOMTR69
C		IMOMTR70
C	DETERMINE THE QUANTIZING FUNCTION	IMOMTR71
C		IMOMTR72
	CALL IEQP01(NOBL,NQUANT,F,IQ,IMIN)	IMOMTR73
C		IMOMTR74
C	NEXT COMPUTE THE QUANTIZED TRANSLATED MEAN FOR EACH ARRAY	IMOMTR75
C		IMOMTR76
	DO 2 I=1,NQUANT	IMOMTR77
	DO 2 J=1,NQUANT	IMOMTR78
	NSI=1	IMOMTR79
	IF (I.NE.1) NSI=IQ(I-1)+2-.MIN	IMOMTR80
	NI=IQ(I)-.MIN +1	IMOMTR81
	NSJ=1	IMOMTR82
	IF (J.NE.1) NSJ=IQ(J-1)+2-.MIN	IMOMTR83
	NFJ=IQ(J) -.MIN +1	IMOMTR84
	IF (NSI.GT.NEI) GO TO 2	IMOMTR85
	IF (NSJ.GT.NEJ) GO TO 2	IMOMTR86
	NADD1=0	IMOMTR87
	NADD2=0	IMOMTR88
	NADD3=0	IMOMTR89
	NADD4=0	IMOMTR90
	DO 9 NI=NSI,NEI	IMOMTR91
	DO 9 NJ=NSJ,NEJ	IMOMTR92
	IJ=INDEX(NI,NJ)	IMOMTR93
	NADD1=NADD1+LEX1(IJ)	IMOMTR94
	NADD2=NADD2+LEX2(IJ)	IMOMTR95
	NADD3=NADD3+LEX3(IJ)	IMOMTR96
	9 NADD4=NADD4+LEX4(IJ)	IMOMTR97
	AMEAN(1)=AMEAN(1)+FLOAT(NADD2*I)	IMOMTR98
	AMEAN(4)=AMEAN(4)+FLOAT(NADD3*I)	IMOMTR99
		IMOMTR00

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```

AMEAN(2)=AMEAN(2)+FLOAT(NADD4*I)
AMEAN(3)=AMEAN(3)+FLOAT(NADD1*I)
2 CONTINUE

```

```

NOW NORMALIZE TO GET THE MEANS

```

```

AMEAN(1)=AMEAN(1)*R2
AMEAN(2)=AMEAN(2)*R4
AMEAN(3)=AMEAN(3)*R1
AMEAN(4)=AMEAN(4)*R3

```

```

NOW DO MOMENT CALCULATIONS

```

```

DO 3 I=1,NQUANT
DO 3 J=1,NQUANT
NSI=1
IF(I.NE.1) NSI=IQ(I-1)+2-IMIN
NEI=IQ(I) -IMIN +1
NSJ=1
IF(J.NE.1) NSJ=IQ(J-1)+2 -IMIN
NFJ=IQ(J) -IMIN +1
IF(NSI.GT.NEI) GO TO 3
IF(NSJ.GT.NEJ) GO TO 3
NADD1=0
NADD2=0
NADD3=0
NADD4=0
DO 13 NI=NSI,NEI
DO 13 NJ=NSJ,NEJ
NINJ=INDEX(NI,NJ)

```

```

SUM UP THE ELEMENTS IN EACH LEX ARRAY

```

```

NADD1=NADD1+LEX1(NINJ)
NADD2=NADD2+LEX2(NINJ)
NADD3=NADD3+LEX3(NINJ)
13 NADD4=NADD4+LEX4(NINJ)

```

```

NORMALIZE

```

```

RL1=FLOAT(NADD1)*R1
RL2=FLOAT(NADD2)*R2
RL3=FLOAT(NADD3)*R3
RL4=FLOAT(NADD4)*R4

```

```

CALCULATE THE MOMENTS

```

```

ANGMOM(1)=ANGMOM(1)+PL2**2
ANGMOM(2)=ANGMOM(2)+PL4**2
ANGMOM(3)=ANGMOM(3)+RL1**2

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```

ANGMOM(4)=ANGMOM(4)+RL3**2
SGMASQ(1)=SGMASQ(1)+((FLOAT(I)-AMEAN(1))**2)*RL2
SGMASQ(2)=SGMASQ(2)+((FLOAT(I)-AMEAN(2))**2)*RL4
SGMASQ(3)=SGMASQ(3)+((FLOAT(I)-AMEAN(3))**2)*RL1
SGMASQ(4)=SGMASQ(4)+((FLOAT(I)-AMEAN(4))**2)*RL3
SGMAXY(1)=SGMAXY(1)+(FLOAT(I)-AMEAN(1))*(FLOAT(J)-AMEAN(1))*RL2
SGMAXY(2)=SGMAXY(2)+(FLOAT(I)-AMEAN(2))*(FLOAT(J)-AMEAN(2))*RL4
SGMAXY(3)=SGMAXY(3)+(FLOAT(I)-AMEAN(3))*(FLOAT(J)-AMEAN(3))*RL1
SGMAXY(4)=SGMAXY(4)+(FLOAT(I)-AMEAN(4))*(FLOAT(J)-AMEAN(4))*RL3
IVDMOM(1)=IVDMOM(1)+RL2/(1.+FLOAT((I-J)**2))
IVDMOM(2)=IVDMOM(2)+RL4/(1.+FLOAT((I-J)**2))
IVDMOM(3)=IVDMOM(3)+RL1/(1.+FLOAT((I-J)**2))
IVDMOM(4)=IVDMOM(4)+RL3/(1.+FLOAT((I-J)**2))
IF(RL2.LT.0.000001)GO TO 50
ENTROP(1)=ENTROP(1)-RL2*A LOG(RL2)
50 IF(RL4.LT.0.000001)GO TO 51
ENTROP(2)=ENTROP(2)-RL4*A LOG(RL4)
51 IF(RL1.LT.0.000001)GO TO 52
ENTROP(3)=ENTROP(3)-RL1*A LOG(RL1)
52 IF(RL3.LT.0.000001)GO TO 53
ENTROP(4)=ENTROP(4)-RL3*A LOG(RL3)
53 CONTINUE
C
C SET UP THE SUM ARRAY
C
K=IABS(I-J)+1
C
C SET UP THE DIFFERENCE ARRAY
C
KS=IABS(I+J)+1
DIF1(K)=DIF1(K)+RL2
DIF2(K)=DIF2(K)+RL4
DIF3(K)=DIF3(K)+RL1
DIF4(K)=DIF4(K)+RL3
SUM1(KS)=SUM1(KS)+RL2
SUM2(KS)=SUM2(KS)+RL4
SUM3(KS)=SUM3(KS)+RL1
SUM4(KS)=SUM4(KS)+RL3
3 CONTINUE
DO 4 I=1,4
4 RATIO(I)=SGMAXY(I)/SGMASQ(I)
C
C CALCULATE THE ENTROPY, AVERAGE, AND THE VARIANCE OF THE DIFFERENCE
C ARRAY
C
DO 31 K=1,NQUANT
IF(DIF1(K).LT.0.000001)GO TO 54
DIFENT(1)=DIFENT(1)-DIF1(K)*A LOG(DIF1(K))
54 IF(DIF2(K).LT.0.000001)GO TO 55
DIFENT(2)=DIFENT(2)-DIF2(K)*A LOG(DIF2(K))

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REPRODUCIBILITY OF THE  
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```

55 IF(DIF3(K).LT.0.000001) GO TO 56
   DIFENT(3)=DIFENT(3)-DIF3(K)*ALOG(DIF3(K))
56 IF(DIF4(K).LT.0.000001) GO TO 57
   DIFENT(4)=DIFENT(4)-DIF4(K)*ALOG(DIF4(K))
57 CONTINUE
   G=FLOAT(K)
   DIFAVE(1)=DIFAVE(1)+(G*DIF1(K))
   DIFAVE(2)=DIFAVE(2)+(G*DIF2(K))
   DIFAVE(3)=DIFAVE(3)+(G*DIF3(K))
   DIFAVE(4)=DIFAVE(4)+(G*DIF4(K))
   DIFVAR(1)=DIFVAR(1)+(G*G)*DIF1(K)
   DIFVAR(2)=DIFVAR(2)+(G*G)*DIF2(K)
   DIFVAR(3)=DIFVAR(3)+(G*G)*DIF3(K)
31  DIFVAR(4)=DIFVAR(4)+(G*G)*DIF4(K)
   DO 32 KK=1,4
32  DIFVAR(KK)=DIFVAR(KK)-(DIFAVE(KK)*DIFAVE(KK))

```

C  
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C

CALCULATE THE ENTROPY, AVERAGE, AND THE VARIANCE OF THE SUM ARRAY

```

DO 33 K=1, NQUAN2
IF(SUM1(K).LT.0.000001) GO TO 58
SUMENT(1)=SUMENT(1)-SUM1(K)*ALOG(SUM1(K))
58 IF(SUM2(K).LT.0.000001) GO TO 59
SUMENT(2)=SUMENT(2)-SUM2(K)*ALOG(SUM2(K))
59 IF(SUM3(K).LT.0.000001) GO TO 60
SUMENT(3)=SUMENT(3)-SUM3(K)*ALOG(SUM3(K))
60 IF(SUM4(K).LT.0.000001) GO TO 61
SUMENT(4)=SUMENT(4)-SUM4(K)*ALOG(SUM4(K))
61 CONTINUE

```

```

G=FLOAT(K)
SUMAVE(1)=SUMAVE(1)+(G*SUM1(K))
SUMAVE(2)=SUMAVE(2)+(G*SUM2(K))
SUMAVE(3)=SUMAVE(3)+(G*SUM3(K))
SUMAVE(4)=SUMAVE(4)+(G*SUM4(K))
SUMVAR(1)=SUMVAR(1)+(G*G)*SUM1(K)
SUMVAR(2)=SUMVAR(2)+(G*G)*SUM2(K)
SUMVAR(3)=SUMVAR(3)+(G*G)*SUM3(K)
33  SUMVAR(4)=SUMVAR(4)+(G*G)*SUM4(K)

```

```

DO 34 KK=1,4
34  SUMVAR(KK)=SUMVAR(KK)-(SUMAVE(KK)*SUMAVE(KK))
RETURN

```

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C

SECTION 11 IMOMTR FOR THE MERGED LEX ARRAY

GET THE PROBABILITY FUNCTION IN F FOR MERGE OPTION

```

911 DO 16 I=1, NOBL
    IA=0

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```

DO 17 J=1,NOBL
IJ=INDEX(I,J)
17 IA=IA+LEX1(IJ)
16 F(I)=FLOAT(IA)/FLOAT(IR1)
C
C FIRST COMPUTE THE TRUE MEAN
C
TMEAN=0
DO 90 I=1,NOBL
90 TMEAN=TMEAN+F(I)*FLOAT(I)
TMEAN=TMEAN+FLOAT(IMIN-1)
C
C SET CUMULATIVE DISTRIBUTION FUNCTION IN F
C
DO 91 I=2,NOBL
91 F(I)=F(I)+F(I-1)
C
C DETERMINE THE QUANTIZING FUNCTION
C
CALL IEQPQ1(NOBL,NQUANT,F,IQ,IMIN)
C
C NEXT COMPUTE THE QUANTIZED TRANSLATED MEAN
C
DO 92 I=1,NQUANT
DO 92 J=1,NQUANT
NSI=1
IF(I.NE.1) NSI=IQ(I-1)+2-IMIN
NEI=IQ(I)-IMIN +1
NSJ=1
IF(J.NE.1) NSJ=IQ(J-1)+2-IMIN
NEJ=IQ(J) -IMIN +1
IF(NSI.GT.NEI) GO TO 92
IF(NSJ.GT.NEJ) GO TO 92
NADD1=0
DO 93 NI=NSI,NEI
DO 93 NJ=NSJ,NEJ
IJ=INDEX(NI,NJ)
93 NADD1=NADD1+LFX1(IJ)
AMEAN(3)=AMEAN(3)+FLOAT(NADD1* I)
92 CONTINUE
C
C NOW NORMALIZE TO GET THE MEANS
C
AMEAN(3)=AMEAN(3)*R1
C
C NOW DO MOMENT CALCULATIONS
C
DO 95 I=1,NQUANT
DO 95 J=1,NQUANT
NSI=1

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IF(I.NE.1) NSI=IQ(I-1)+2-.MIN
NEI=IQ(I) -IMIN +1
NSJ=1
IF(J.NE.1) NSJ=IQ(J-1)+2 -IMIN
NEJ=IQ(J) -IMIN +1
IF(NSI.GT.NEI) GO TO 95
IF(NSJ.GT.NEJ) GO TO 95
NADD1=0
DO 96 NI=NSI,NEI
DO 96 NJ=NSJ,NEJ
NINJ=INDEX(NI,NJ)

C      SUM UP THE ARRAY
C
C      NADD1=NADD1+LEX1(NINJ)
C
C      NORMALIZE
C
C      RL1=FLOAT(NADD1)*R1
C
C      COMPUTE MOMENTS
C
ANGMOM(3)=ANGMOM(3)+R_1**2
SGMASQ(3)=SGMASQ(3)+((FLOAT(I)-AMEAN(3))**2)*RL1
SGMAXY(3)=SGMAXY(3)+(FLOAT(I)-AMEAN(3))*(FLOAT(J)-AMEAN(3))*RL1
IVDMOM(3)=IVDMOM(3)+RL1/(1.+FLOAT((I-J)**2))
IF(RL1.LT.0.000001)GO TO 533
ENTROP(3)=ENTROP(3)-RL1*A LOG(RL1)
533  CONTINUE

C      SET UP THE SUM ARRAY
C
C      K=IABS(I-J)+1
C
C      SET UP THE DIFFERENCE ARRAY
C
KS=IABS(I+J)+1
DIF3(K)=DIF3(K)+RL1
SUM3(KS)=SUM3(KS)+RL1
95   CONTINUE
RATIO(3)=SGMAXY(3)/SGMASQ(3)

C      CALCULATE THE ENTROPY,AVERAGE,AND THE VARIANCE OF THE DIFFERENCE
C      ARRAY
C
DO 97 K=1,NQUANT
IF(DIF3(K).LT.0.000001) GO TO 577
DIFFNT(3)=DIFFNT(3)-DIF3(K)*A LOG(DIF3(K))
577  CONTINUE
G=FLOAT(K)

```

IMOMTR01  
 IMOMTR02  
 IMOMTR03  
 IMOMTR04  
 IMOMTR05  
 IMOMTR06  
 IMOMTR07  
 IMOMTR08  
 IMOMTR09  
 IMOMTR10  
 IMOMTR11  
 IMOMTR12  
 IMOMTR13  
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 IMOMTR36  
 IMOMTR37  
 IMOMTR38  
 IMOMTR39  
 IMOMTR40  
 IMOMTR41  
 IMOMTR42  
 IMOMTR43  
 IMOMTR44  
 IMOMTR45  
 IMOMTR46  
 IMOMTR47  
 IMOMTR48  
 IMOMTR49  
 IMOMTR50

REPRODUCIBILITY OF THE  
 ORIGINAL PAGE IS POOR

	JIFAVE(3)=DIFAVE(3)+(G*DIF3(K))	IMOMTR5
97	DIFVAR(3)=DIFVAR(3)+(G*G)*DIF3(K)	IMOMTR5
	D.FVAR(3)=DIFVAR(3)-(DIFAVE(3)*DIFAVE(3))	IMOMTR5
C		IMOMTR5
C	CALCULATE THE ENTROPY, AVERAGE, AND THE VARIANCE OF THE SUM	IMOMTR5
C	ARRAY	IMOMTR5
	DO 98 K=1,NQUAN?	IMOMTR5
	IF(SUM3(K).LT.0.000001) GO TO 99	IMOMTR5
	SJMENT(3)=SUMENT(3)-SUM3(K)*ALOG(SUM3(K))	IMOMTR6
99	CONTINUE	IMOMTR6
	G=FLOAT(K)	IMOMTR6
	SUMAVE(3)=SUMAVE(3)+(G*SUM3(K))	IMOMTR6
98	SUMVAR(3)=SUMVAR(3)+(G*G)*SUM3(K)	IMOMTR6
	SUMVAR(3)=SUMVAR(3)-(SUMAVE(3)*SUMAVE(3))	IMOMTR6
	RETURN	IMOMTR6
	END	IMOMTR6

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

CCOR

SUBROUTINE COR

WRITTEN BY SAM SHANMUGAM

NOV 1972

COR00001  
COR00002  
COR00003  
COR00004  
COR00005  
COR00006  
COR00007  
COR00008  
COR00009  
COR00010  
COR00011  
COR00012  
COR00013  
COR00014  
COR00015  
COR00016  
COR00017  
COR00018  
COR00019  
COR00020  
COR00021  
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COR00031  
COR00032  
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COR00036  
COR00037  
COR00038  
COR00039  
COR00040  
COR00041  
COR00042  
COR00043  
COR00044  
COR00045

\*\*\*\*\*  
OBJECTIVE.  
-----

THIS PROGRAM CALCULATES THREE MEASURES OF  
CORRELATION COR1, COR2, COR3, BETWEEN TWO  
DISCRETE RANDOM VARIABLES X AND Y WHOSE  
JOINT PROBABILITIES OF OCCURANCE ARE  
STORED IN THE ARRAY PXY

ENTRY POINT.  
-----

CALL COR(PXY, N, IOPT, Q, COR1, COR2, COR3)

INPUT ARGUMENTS.  
-----

PXY      ARRAY OF JOINT PROBABILITIES  
N        SIZE OF THE ARRAY PXY  
IOPT     OPTION FLAG--IF IOPT=0 THEN COR1 AND  
          COR2 ONLY WILL BE COMPUTED. IF  
          IOPT=1, THEN COR3 WILL BE COMPUTED  
          SCRATCH ARRAY OF SIZE N X N. THIS  
          ARRAY IS NEEDED ONLY IF IOPT IS  
          NON-ZERO. IF IOPT IS ZERO THEN A  
          DUMMY VARIABLE MAY BE SUBSTITUTED FOR  
          THE ARGUMENT Q

OUTPUT ARGUMENTS.  
-----

COR1     MAXIMAL CORRELATION MEASURE  
COR2     INFORMATION MEASURE OF CORRELATION  
COR3     SECOND TYPE OF MAXIMAL MEASURE

BIBLIOGRAPHY.  
-----

ANNALS OF MATHEMATICAL STATISTICS, VOL. 43, 1962,  
P. 587, "MUTUAL INFORMATION AND MAXIMAL  
CORRELATION AS MEASURE OF DEPENDENCE" BY  
C.B. BELL.

CAUTION.

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

CC

1. THE ARRAY PXY MUST HAVE A DIMENSION OF  $N \times N$  IN THE CALLING PROG. OR PXY SHOULD BE A ONE DIMENSIONAL VECTOR, CONTAINING JOINT PROBABILITIES IN A COLUMN BY COLUMN ARRANGEMENT.  
 2. IF N IS LARGE, THE COMPUTATIONS FOR COR3 WILL TAKE CONSIDERABLE TIME. HENCE THE USE OF THIS ROUTINE IS RESTRICTED TO N LESS THAN OR EQUAL TO 32

-----  
 COMPUTATIONS.  
 -----

```

PX(I) = SUM PXY(I,J)
PY(J) = SUM PXY(I,J)
HXY = SUM SUM LOG(PXY(I,J) * PXY(I,J))
HXY1 = SUM SUM LOG((PX(I)*PY(J)) * PXY(I,J))
HXY2 = SUM SUM (LOG(PX(I)*PY(J)) * PX(I)*PY(J))
HX = SUM (LOG(PX(I)) * PX(I))
HY = SUM (LOG(PY(J)) * PY(J))
R = HXY2 - HXY . EMAX = MAX(HX, HY)
COR1 = (HXY - HXY1) / EMAX
COR2 = SQRT(1.0 - EXP(-2.5 * R))
COR3 = IS COMPUTED USING THE EIGEN VECTOR
CORRESPONDING TO THE SECOND LARGEST
EIGEN VALUE OF Q*QTRANSPOSE, WHERE
Q(I,J) = PXY(I,J) / SQRT(PX(I)*PY(J))
    
```

REPRODUCIBILITY OF THE  
 ORIGINAL PAGE IS POOR

\*\*\*\*\*  
 SUBROUTINE COR(PXY,N,IOP,Q,COR1,COR2,COR3)

DIMENSION PXY(1),Q(1)  
 DIMENSION PX(64),PY(64),E(64),V(128),B(64),C(64),D(64),F(64)  
 DIMENSION IZERO(32)  
 \*\*\*\*\*

INITIALIZE PX,PY,HX,HY,HXY,HXY1,AND HXY2

```

DO 80 I=1,N
PY(I)=0.0
80 PY(I)=0.0
    
```

COR00051  
 COR00052  
 COR00053  
 COR00054  
 COR00055  
 COR00056  
 COR00057  
 COR00058  
 COR00059  
 COR00060  
 COR00061  
 COR00062  
 COR00063  
 COR00064  
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 COR00086  
 COR00087  
 COR00088  
 COR00089  
 COR00090  
 COR00091  
 COR00092  
 COR00093  
 COR00094  
 COR00095  
 COR00096  
 COR00097  
 COR00098  
 COR00099  
 COR00100

HX=0.0  
 HY=0.0  
 HXY=0.0  
 HXY1=0.0  
 HXY2=0.0

COMPUTE THE MARGINALS AND THEIR ENTROPY

C  
 C  
 C

DO 82 I=1,N  
 DO 81 J=1,N  
 IJ=(J-1)\*N+I  
 81 PX(I)=PX(I)+PXY(IJ)  
 IF(PX(I).LT.0.000001) GO TO 82  
 HX=HX-(ALOG(PX(I)))\*(PX(I))  
 82 CONTINUE  
 DO 84 J=1,N  
 DO 83 I=1,N  
 IJ=(J-1)\*N+I  
 83 PY(J)=PY(J)+PXY(IJ)  
 IF(PY(J).LT.0.000001) GO TO 84  
 HY=HY-ALOG(PY(J))\*PY(J)  
 84 CONTINUE

COMPUTE THE ENTROPY OF THE JOINT DISTR.

C  
 C  
 C

DO 69 I=1,N  
 DO 69 J=1,N  
 IJ=(J-1)\*N+I  
 IF(PXY(IJ).LT.0.000001) GO TO 68  
 HXY=HXY-(ALOG(PXY(IJ)))\*(PXY(IJ))  
 68 PXPY=PX(I)\*PY(J)  
 IF(PXPY.LT.0.000001) GO TO 69  
 HXY1=HXY1-(ALOG(PXPY))\*PXY(IJ)  
 HXY2=HXY2-(ALOG(PXPY))\*PXPY  
 69 CONTINUE

COMPUTE COR1 AND COR2

C  
 C  
 C

EMAX=HX  
 IF(HX.LT.-Y) EMAX=HY  
 COR1=(HXY-HXY1)/EMAX  
 R=HXY2-HXY  
 COR2 = SQRT(1.0-EXP(-2.0\*R))

IF COR3 NOT ASKED FOR RETURN

C  
 C  
 C

IF(IOPT.EQ.0) RETURN

SCAN PXY AND DELETE ROWS OF ZEROS  
 AND COLUMNS OF ZEROS

C  
 C

DO 599 INDFX=1,32

COR00101  
 COR00102  
 COR00103  
 COR00104  
 COR00105  
 COR00106  
 COR00107  
 COR00108  
 COR00109  
 COR00110  
 COR00111  
 COR00112  
 COR00113  
 COR00114  
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 COR00117  
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 COR00119  
 COR00120  
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 COR00148  
 COR00149  
 COR00150

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599	IZERO(INDFX)=0	COR00151
	NZERO=0	COR00152
	DO 600 I=1,N	COR00153
	I.=I	COR00154
	IF(PX(I).GT.0.000001) GO TO 601	COR00155
	NZERO=NZERO+1	COR00156
	IZERO(NZERO)=II	COR00157
601	CONTINUE	COR00158
600	CONTINUE	COR00159
	IF(NZERO.EQ.0) GO TO 651	COR00160
	JJ=0	COR00161
	DO 650 J=1,N	COR00162
	DO 650 I=1,N	COR00163
	DO 640 KK=1,NZERO	COR00164
	NDEX=IZERO(KK)	COR00165
	IF((I.EQ.NDEX).OR.(J.EQ.NDEX)) GO TO 649	COR00166
640	CONTINUE	COR00167
	JJ=JJ+1	COR00168
	IJ=(J-1)*N+I	COR00169
	PXY(JJ)=PXY(IJ)	COR00170
649	CONTINUE	COR00171
650	CONTINUE	COR00172
651	CONTINUE	COR00173
	REMOVE ZERO ENTRIES IN THE MARGINALS	COR00174
C	JJ=0	COR00175
	DO 661 I=1,N	COR00176
	IF(PX(I).LT.0.000001) GO TO 662	COR00177
	JJ=JJ+1	COR00178
	PX(JJ)=PX(I)	COR00179
662	CONTINUE	COR00180
661	CONTINUE	COR00181
	JJ=0	COR00182
	DO 671 I=1,N	COR00183
	IF(PY(I).LT.0.000001) GO TO 672	COR00184
	JJ=JJ+1	COR00185
	PY(JJ)=PY(I)	COR00186
672	CONTINUE	COR00187
671	CONTINUE	COR00188
	NNNN=N	COR00189
	N=N-NZERO	COR00190
C		COR00191
C		COR00192
C		COR00193
C		COR00194
C		COR00195
	DO 58 I=1,N	COR00196
	DO 58 J=1,N	COR00197
	IJ=(J-1)*N+I	COR00198
	CONS=SQRT(PY(J))	COR00199
58	Q(IJ)=PXY(IJ)/CONS	COR00200

REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR

C  
C  
C  
C

COMPUTE THE UPPER/DIAG ELEMENTS OF Q\*QT  
STORE IN Q

```

DO 49 I=1,N
DO 51 J=I,N
B(J)=0.0
DO 52 K=1,N
IK=(K-1)*I+1
JK=(K-1)*N+J
52 B(J)=B(J)+Q(IK)*Q(JK)
51 CONTINUE
DO 50 J=1,N
IJ=(J-1)*N+1
50 Q(IJ)=B(J)
49 CONTINUE
    
```

C  
C  
C

FILL IN THE BELOW DIAG ELEMENTS OF Q\*QT

```

DO 48 J=1,N
DO 48 I=J,N
IJ=(J-1)*N+I
JI=(I-1)*N+J
48 Q(IJ)=Q(JI)
    
```

C  
C

FORM SQRT(PX) \* Q\*QT \* SQRT(PY)  
STORE IN Q

```

DO 91 I=1,N
DO 91 J=1,N
IJ=(J-1)*N+I
91 Q(IJ)=Q(IJ)/(SQRT(PX(I))*PX(J))
    
```

C  
C  
C  
C  
C  
C  
C  
C  
C  
C  
C  
C  
C  
C  
C  
C

GET THE EIGEN VECTORS AND EIGEN VALUES  
OF Q\*QT

CALL THE SUBROUTINE TO GET THE EIGENVALS  
GET A MAX OF 5 EIGENVALUES. IF ALL FIVE  
ARE NEAR UNITY, SET COR3=0.9999. RETURN  
IF ALL OF THEM (OTHER THAN THE FIRST ONE)  
ARE LESS THAN 0.001, SET COR3=0.0001 AND  
RETURN. THE EIGEN VALUES ARE CALCULATED  
WITH AN ACCURACY OF 0.0001.

```

MAX=5
IF(N.LT.MAX) MAX=N
CR=0.0001
EPS=0.00001
CALL SFA02D(Q,N,N,CR,EPS,MAX,NE,E,V,B,C,D,F,IE)
    
```

C

COR00201  
COR00202  
COR00203  
COR00204  
COR00205  
COR00206  
COR00207  
COR00208  
COR00209  
COR00210  
COR00211  
COR00212  
COR00213  
COR00214  
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COR00216  
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COR00221  
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COR00245  
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COR00247  
COR00248  
COR00249  
COR00250



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20.860

SUBROUTINE COP

	IF(NE.GT.1) GO TO 60	COR0025
C		COR0025
C	SET COR3=0.0001 AND RETURN	COR0025
C		COR0025
	COR3=0.01	COR0025
	WRITE(6,200)	COR0025
200	FORMAT(1F1,10X,' ALL EIGEN VALUES OTHER THAN THE FIRST ARE LESS THAN 0.001. COR3 IS SET=0.0001')/10X,'ABORT AA')	COR0025
	N=NNNN	COR0025
	RETURN	COR0026
60	CONTINUE	COR0026
C	FIND THE EIGEN VALUE CLOSEST TO 1.0	COR0026
	DO 63 I=1,NE	COR0026
	DIF=ABS(E(I)-1.0)	COR0026
	JJE=I	COR0026
	IF (DIF.GT.0.00003) GO TO 64	COR0026
63	CONTINUE	COR0026
C		COR0026
C	IF PROG IS AT THIS POINT, THEN THE FIRST FIVE EIGEN VALUES ARE CLOSE TO 1.0. SET COR3=0.9999 AND RETURN	COR0027
C		COR0027
C		COR0027
C		COR0027
	COR3=0.9999	COR0027
	WRITE(6,202)	COR0027
202	FORMAT(1H1,10X,' THE FIRST 5 EIGEN VALUES ARE NEAR UNITY. '/10X,' COR3 IS SET =0.999.....ABORT BB')	COR0027
	1 COR3 IS SET =0.999.....ABORT BB')	COR0027
	N=NNNN	COR0027
	RETURN	COR0027
C		COR0027
C	FOUND A PROPER EIGEN VALUE	COR0028
64	IFOUND=JJE	COR0028
	COR3=SQRT(E(IFOUND))	COR0028
	N=NNNN	COR0028
	RETURN	COR0028
	END	COR0028

```

CIEQPQ1                                I-E-Q-P-Q-1                                IEQPQ101
C                                          WRITTEN BY DENISH GOEL                                SEPT 1971                                IEQPQ102
C                                          DESCRIPTION OF PROGRAM.                                IEQPQ103
C                                          THIS SUBROUTINE DETERMINES K LEVELS OF QUANTIZING FOR AN ARRAY
C                                          FOR WHICH THE CUMMULATIVE DISTRIBUTION FUNCTION OF ALL THE EL-
C                                          MFMENTS HAVE ALREADY BEEN OBTAINED.                                IEQPQ104
C                                          ENTRY POINT.                                IEQPQ105
C                                          CALL IEQPQ1(N,K,F,IQ,IMIN)                                IEQPQ106
C                                          ARGUMENTS.                                IEQPQ107
C                                          N                                NUMBER OF ITEMS TO BE QUANTIZED, THE
C                                          DIMENSION OF THE F ARRAY.                                IEQPQ108
C                                          K                                THE NUMBER OF QUANTIZING LEVELS                                IEQPQ109
C                                          F                                INPUT ARRAY OF CUMULATIVE DISTRIBUTION
C                                          FUNCTION.                                IFQPQ110
C                                          IQ                                OUTPUT ARRAY OF QUANTIZING LEVELS.                                IEQPQ111
C                                          IMIN                                THE LOWEST POSSIBLE LEVEL IN
C                                          THE INPUT DATA.                                IEQPQ112
C                                          SUBROUTINE IEQPQ1(N,K,F,IQ,IMIN)                                IEQPQ113
C                                          DIMENSION F(1),IQ(1)                                IEQPQ114
C                                          DIF=10.**6                                IEQPQ115
C                                          OBTAIN THE FIRST QUANTIZING LEVEL.                                IEQPQ116
C                                          GO THRU THE WHOLE ARRAY OF C.D.F,S                                IEQPQ117
C                                          DO 1 J=1,N                                IEQPQ118
C                                          FIND PERCENTAGE OF DISTRIBUTION FOR FIRST QUANTIZING LEVEL AND
C                                          CHECK FOR THE NFAREST C.J.F.                                IEQPQ119
C                                          X=ABS(1./FLOAT(K)-F(J))                                IEQPQ120
C                                          IF(DIF.LE.X)GO TO 1                                IEQPQ121
C                                          DIF=X                                IEQPQ122
C                                          ISAVE=J                                IEQPQ123
C                                          1 CONTINUE                                IEQPQ124
C                                          FIRST QUANTIZING LEVEL                                IEQPQ125
C                                          IQ(1)=ISAVE+IMIN-1                                IEQPQ126
C                                          TO GO FOR NEXT LEVEL                                IEQPQ127
C                                          IEQPQ128
C                                          IEQPQ129
C                                          IEQPQ130
C                                          IEQPQ131
C                                          IEQPQ132
C                                          IEQPQ133
C                                          IEQPQ134
C                                          IEQPQ135
C                                          IEQPQ136
C                                          IEQPQ137
C                                          IEQPQ138
C                                          IEQPQ139
C                                          IEQPQ140
C                                          IEQPQ141
C                                          IEQPQ142
C                                          IEQPQ143
C                                          IEQPQ144
C                                          IEQPQ145
C                                          IEQPQ146
C                                          IEQPQ147
C                                          IEQPQ148
C                                          IEQPQ149
C                                          IEQPQ150

```

```

LFTOFF=ISAVE
DO 2 I=2,K
C
C   DECIDE FOR OTHER QUANTIZING LEVELS IN THE SIMILAR WAY.
C
DIF=10.**6
DO 3 J=LFTOFF,N
C
C   THE PERCENTAGE OF DISTRIBUTION FOR NEXT QUANTIZING LEVEL WILL
C   BE DECIDED AMONG REST OF ELEMENTS.
C
X=ABS(((1.-F(LFTOFF))/FLOAT(K-I+1))+F(LFTOFF)-F(J))
IF (DIF.LE.X) GO TO 3
DIF=X
ISAVE=J
C
C   GET THE NEXT QUANTIZING LEVEL
C
3 CONTINUE
IQ(I)=ISAVE+IMIN-1
LFTOFF=ISAVE
2 CONTINUE
RETURN
END

```

```

IEQPQ151
IEQPQ152
IEQPQ153
IEQPQ154
IEQPQ155
IEQPQ156
IEQPQ157
IEQPQ158
IEQPQ159
IEQPQ160
IEQPQ161
IEQPQ162
IEQPQ163
IEQPQ164
IEQPQ165
IEQPQ166
IEQPQ167
IEQPQ168
IEQPQ169
IEQPQ170
IEQPQ171
IEQPQ172
IEQPQ173
IEQPQ174

```

CRITOWT	R-I-T-O-W-T	JUNE 1973	RITOWT01
C			RITOWT02
C	WRITTEN BY RMH	SEPT 1971	RITOWT03
C	VERSION 1 BY RJ BOSLEY FOR TAPE OUTPUT OPTION	NOV 1972	RITOWT04
C	VERSION 2 BY RJ BOSLEY FOR MERGE	JUNE 1973	RITOWT05
C			RITOWT06
C	DESCRIPTION OF PROGRAM.		RITOWT07
C	THIS SUBROUTINE PUNCHES, PRINTS, OR WRITES TO FILE (IF) THE TEXTURE		RITOWT08
C	DATA ACCORDING TO THE PNCH / MERGE OPTIONS		RITOWT09
C			RITOWT10
C	ENTRY POINT.		RITOWT11
C	CALL RITOWT(LEX1, LEX2, LEX3, LEX4, G, IQ, MERGE, MERGE, IF, PICTUR)		RITOWT12
C			RITOWT13
C	ARGUMENTS.		RITOWT14
C	LEX1-LEX4	ADDRESS INDEXS FOR THE LEX ARRAYS	RITOWT15
C	G	CDF FOR THE IMAGE DATA	RITOWT16
C	IQ	QUANTIZED OUTPUT OF IEQPQ1 OF NQUANT	RITOWT17
C		LEVELS	RITOWT18
C	MERGE	OPTION TO MERGE THE FOUR LEX ARRAYS	RITOWT19
C		INTO ONE ARRAY	RITOWT20
C	IF	FILE CODE FOR OUTPUT TAPE	RITOWT21
C	PICTUR	OPTION TO PRINT A PICTURE OF THE	RITOWT22
C		IMAGE, USED TO VARY SPACING	RITOWT23
C			RITOWT24
C	INTERNAL PARAMETERS.		RITOWT25
C	PNCH	=TAPE FOR TAPE OUTPUT ON FILE IF	RITOWT26
C	PNCH	=Y FOR PUNCH OUTPUT	RITOWT27
C	PNCH	=N FOR PRINTER OUTPUT ONLY	RITOWT28
C	NPED	REDUCTION FACTOR FROM FPLXIT	RITOWT29
C	NLAYER	THE POWER TO WHICH NRED IS RAISED	RITOWT30
C	NFT	AMOUNT OF REDUCTION IN FPLXIT	RITOWT31
C	N	THE NUMBER OF ANGLES USED IN LEX	RITOWT32
C		ARRAYS	RITOWT33
C	M	THE LEX ARRAY CONTAINING ALL THE	RITOWT34
C		MERGED ARRAYS	RITOWT35
C	M1	IMAGE ROW INDEX	RITOWT36
C	N1	IMAGE COLUMN INDEX	RITOWT37
C	KOUNT	CARD COUNTER	RITOWT38
C	ANGMOM...CORMAX	FEATURES	RITOWT39
C	NOBL	NUMBER OF BRIGHTNESS LEVELS IN IMAGE	RITOWT40
C	IMT	LINE COUNTER FOR MERGE OPTION	RITOWT41
C			RITOWT42
C	SUBROUTINE RITOWT(LEX1, LEX2, LEX3, LEX4, G, IQ, MERGE, IF, PICTUR)		RITOWT43
C			RITOWT44
C	DIMENSION LEX1(1), LEX2(1), LEX3(1), LEX4(1), G(64), IQ(64), B(4)		RITOWT45
C	COMMON M1, N1, TYPE, F(14), ZDD(9), ANGMOM(4), AMEAN(4), SGMASQ(4),		RITOWT46
C	1 SGMAXY(4), DIFMOM(4), RATIO(4), VIDMOM(4), THEAN, LEAST1, NRED		RITOWT47
C	COMMON NLAYER, NSTART, NTIMES, NO, PNCH		RITOWT48
C	COMMON /E./ENTROP(4), DIFENT(4), DIFAVE(4), DIFVAR(4), SUMENT(4),		RITOWT49
C	1SUMAVE(4), SUMVAR(4)		RITOWT50

```

COMMON /CORREL/CORINF(4),CORMUT(4),CORMAX(4)
LOGICAL MERGE,PICTUR
DATA B(1),B(2),B(3),B(4)/0.,45.,90.,135./
DATA TAPE/1HT/
DATA Y/1HY/
DATA KOUNT/0/,IMT/0/
NFT=NREJ**NLAYER
N=4
KKJ=50
M=3

```

C  
C  
C  
C  
C  
C  
C

CHECK FOR THE MERGE OPTION

REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR

IF (MERGE) GO TO 22

PUNCH OR NO PUNCH

IF (PNCH.NE.Y) GO TO 400

PUNCH TEXTURE FEATURES FOR ALL FOUR ANGLES

```

106 WRITE(43,60) M1,N1,NFT
   FORMAT(1X,2A5,12HIS COMPLETED)
   KOUNT=KOUNT+1
   WRITE(43,600) (ANGMOM(K),K=1,4), (ENTROP(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (RATIO(K),K=1,4), (SGMASQ(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (SGMAXY(K),K=1,4), (AMEAN(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (VIDMOM(K),K=1,4), (TMEAN,K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (DIFENT(K),K=1,4), (DIFAVE(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (DIFVAR(K),K=1,4), (SUMENT(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (SUMAVE(K),K=1,4), (SUMVAR(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,600) (CORINF(K),K=1,4), (CORMUT(K),K=1,4), KOUNT
   KOUNT=KOUNT+1
   WRITE(43,601) (CORMAX(K),K=1,4), KOUNT
600 FORMAT(1X,8F9.5,I7)
601 FORMAT(1X,4F9.4,38X,I5)
661 FORMAT(1X,4F9.5,36X,I7)
400 CONTINUE

```

C  
C  
C

WRITE TEXTURE FEATURES TO TAPE FILE 'IF'

IF (PNCH.NE.TAPE) GO TO 500

WRITE(IF) M1,N1,NFT, (ANGMOM(K),K=1,N), (ENTROP(K),K=1,N), (RATIO(K), K=1,N), (SGMASQ(K),K=1,N), (SGMAXY(K),K=1,N), (AMEAN(K),K=1,N), (VIDMOM(K),K=1,N), (TMEAN(K),K=1,N), (DIFENT(K),K=1,N), (DIFAVE(K),K=1,N), (DIFVAR(K),K=1,N), (SUMENT(K),K=1,N), (SUMAVE(K),K=1,N), (SUMVAR(K),K=1,N), (CORINF(K),K=1,N), (CORMUT(K),K=1,N), (CORMAX(K),K=1,N)

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RITOWT99  
RITOWT00

```

1K=1,N), (SGMASQ(K),K=1,N), (SGMAXY(K),K=1,N), (AMEAN(K),K=1,N),
1(VIOMOM(K),K=1,N), (TMFAN,K=1,N), (DIFENT(K),K=1,N),
1(IFAVF(K),K=1,N), (DIFVAR(K),K=1,N), (SUMFNT(K),K=1,N),
1(SUMAVE(K),K=1,N), (SUMVAR(K),K=1,N), (CORINF(K),K=1,N), (CORMUT(K),
1K=1,N), (CORMAX(K),K=1,N)
500 CONTINUE
C
C PRINT TEXTURE FEATURES FOR EACH ANGLE AND TITLE
C
WRITE(6,60) M1,N1,NFT
60 FORMAT(/' THE SCENE (' ,I2,',',I2,') HAS BEEN REDUCED BY ',I5)
WRITE(6,303)
303 FORMAT(6H ANGLE ,9H ANGMOM ,8H ENTROP ,8H RATIO ,8H SGMASQ ,
18H SGMAXY ,8H IVDMOM ,8H DIFENT ,8H DIFAVE ,8H DIFVAR ,8H SUMENT ,
18H SUMAVE ,8H SUMVAR ,8H CORINF ,7HCOPMUT ,8H CORMAX )
WRITE(6,300) (B(K), ANGMOM(K), ENTROP(K), RATLO(K), SGMASQ(K), SGMAXY(K),
1,VIOMOM(K), DIFENT(K), DIFAVE(K), DIFVAR(K), SUMENT(K), SUMAVE(K),
2SUMVAR(K), CORINF(K), CORMUT(K), CORMAX(K), K=1,N)
300 FORMAT(1X,F5.1,15F8.4)
WRITE(6,600) TMEAN
100 CONTINUE
C
C IF NEITHER PNCH NOR TAPE, PRINT LEX ARRAYS
C
IF((PNCH.EQ.Y).OR.(PNCH.EQ.TAPE)) RETURN
WRITE(6,30) G
WRITE(6,31) IQ
30 FORMAT(2H F/(1X,16F7.3))
31 FORMAT(3H IQ/(1X,16I7))
NOBL=IDD(1)-IDD(2)+1
C
C IF MERGE, JUST DO LEX1 AND RETURN
C
IF(MERGE) GO TO 54
WRITE(6,566)
566 FORMAT(/10X,9H0 DEGREES)
DO 50 I=1,NOBL
NS=1*(I-1)/2+1
NE=(I+1)*1/2
C
C PRINT LEX2 FOR 0 DEGREES
C
50 WRITE(6,700) (LFX2(J),J=NS,NE)
700 FORMAT(1X,26I5)
WRITE(6,567)
567 FORMAT(/10X,10H45 DEGREES)
DO 51 I=1,NOBL
NS=1*(I-1)/2+1
NE=(I+1)*1/2
C

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RITOWT50

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

```

C      PRINT LEX4 FOR 45 DEGREES
C
51 WRITE(6,700) (LEX4(J),J=NS,NE)
   WRITE(6,99)
99  FORMAT(1F1)
54   WRITE(6,568)
568 FORMAT(//10X,10490 DEGREES)
   DO 52 I=1,NOBL
   NS=I*(I-1)/2+1
   NF=(I+1)*I/2

C      PRINT LEX1 FOR 90 DEGREES
C
52 WRITE(6,700) (LEX1(J),J=NS,NE)
   IF(MERGE) RETURN
   WRITE(6,569)
569 FORMAT(//10X,114135 DEGREES)
   DO 53 I=1,NOBL
   NS=I*(I-1)/2+1
   NF=(I+1)*I/2

C      PRINT LEX3 FOR 135 DEGREES
C
53 WRITE(6,700) (LEX3(J),J=NS,NE)
   RETURN

C      RITOWT FOR THE MERGE OPTION
C
C      CHECK TO SEE IF A PICTURE HAS BEEN PRINTED
C
22  IF(PICTUR) GO TO 23
C
C      INCREMENT PAGE COUNT
C
   IMT=IMT+1

C      IF PAGE IS FULL GO TO TOP OF NEXT PAGE AND WRITE TITLE
C
   IF(IMT.LE.1) WRITE(6,662)
   IF(IMT.GE.14) IMT=0
662  FORMAT(1F1,40X,'ERTS TEXTURE ANALYSIS'//
1     1X,'ANGLE ANGMOM ENTROP RATIO SGMA2 SGMAXY IVDORITOWT9
1M DIFENT DIFAVE DIFVAR SUMENT SUMAVE SUMVAR COPINF CORNUT RITOWT9
1 CORMAX *//)

C      CHECK FOR PUNCH
C
23  IF(PNCH.NE.Y) GO TO 40
C
C      PUNCH THE MERGED TEXTURE FEATURES

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ORIGINAL PAGE IS POOR

```

C
WRITE(43,663) M1,N1,NFT,ANGMOM(M),ENTROP(M),RATIO(M),SGMASQ(M),
1 SGMAXY(M),AMEAN(M),VIDMOM(M),KOUNT
KOUNT=KOUNT+1
WRITE(43,664) TMEAN,DIFENT(M),DIFAVE(M),DIFVAR(M),SUMENT(M),
1 SUMAVE(M),SUMVAR(M),KOUNT
KOUNT=KOUNT+1
WRITE(43,665) CORINF(M),CORMUT(M),CORMAX(M),KOUNT
KOUNT=KOUNT+1
663 FORMAT(12,1X,I2,I2,1X,7F9.5,5)
664 FORMAT(8X,7F9.5,I9)
665 FORMAT(8X,3F9.5,36X,I9)
40 CONTINUE
C CHECK FOR TAPE OUTPUT
IF(PNCH.NE.TAPE) GO TO 41

C
C WRITE OUT ONTO FILE 'IF' THE MERGED TEXTURE FEATURES
C
WRITE(IF) M1,N1,NFT,ANGMOM(M),ENTROP(M),RATIO(M),SGMASQ(M),
1 SGMAXY(M),AMFAN(M),VIDMOM(M),TMEAN,DIFENT(M),DIFAVE(M),
2 DIFVAR(M),SUMENT(M),SUMAVE(M),SUMVAR(M),CORINF(M),CORMUT(M),
3 CORMAX(M)

C
C IN ANY CASE,PRINT THE MERGED TEXTURE FEATURES
C
41 WRITE(6,60) M1,N1,NFT
WRITE(6,666) ANGMOM(M),ENTROP(M),RATIO(M),SGMASQ(M),SGMAXY(M)
1,VIDMOM(M),DIFENT(M),DIFAVE(M),DIFVAR(M),SUMENT(M),SUMAVE(M),
2SUMVAR(M),CORINF(M),CORMUT(M),CORMAX(M)
666 FORMAT(1X,'MERGE',15F8.4)
WRITE(6,667) TMEAN
667 FORMAT(1X,F9.5/)

C
C NOW GO PRINT OUT THE MERGED _EX1 ARRAY AND RETURN
C
GO TO 100
END
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REPRODUCIBILITY OF THE  
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## IV.2-c Cross-Band Texture Analysis Program Listings

SPECTR

GETIM / GETIT

ERTS (see IV.2-a)

DIFFER

COVAR

MNCVIN / MNCV

CORREL

CSPECTR

CROSS-BAND TEXTURE ANALYSIS

JAN 1974

WRITTEN BY RJ BOSLEY

DESCRIPTION OF PROGRAM

THIS PROGRAM IS THE MAINLINE OF PROGRAMS WHICH OBTAIN A NUMLIN X NUMPPL X NDIM SUBIMAGE FROM THE ERTS INPUT TAPE AND FOR EACH SUBIMAGE CALCULATES THE COVARIANCE MATRIX AND THE CORRELATION MATRIX. THIS MATRIX IS WRITTEN TO FILE IFILE FOR FURTHER ANALYSIS.

THE ERTS TAPE IS PROCESSED IN HORIZONTAL ROWS OF SUBIMAGES. SUBIMAGES MAY OVERLAP HORIZONTALLY AND VERTICALLY, AND THE DISTANCE BETWEEN NEIGHBORING CELLS USED IN THE DIFFERENCE ARRAY IS VARIABLE.

NOTE---

ERTS INPUT TAPE MUST BE ON FILE ES.  
A RANDOM DISC FILE IS REQUIRED ON FILE 11--- \$ FILE 11,A11R,30R

INTERNAL PARAMETERS

- NDIM NUMBER OF GREY TONE N-TUPLE COMP.
- NUMLIN NUMBER OF LINES IN A SUBIMAGE
- NUMPPL NUMBER OF COLUMNS IN A SUBIMAGE
- NROW ROW COORD FOR THE SUBIMAGE
- NCOL COLUMN COORD FOR THE SUBIMAGE
- IDIST DISTANCE BETWEEN NEIGHBORING RES. CELLS FOR THE DIFFERENCE IMAGE
- IRSTRT STARTING ROW FOR THIS RUN
- IRSTOP STOPPING ROW FOR THIS RUN
- SET=0, THE STRIP WILL BE PROCESSED TO ITS END OF FILE
- LAPHOR HORIZONTAL OVERLAP OF SUBIMAGES
- LAPVER VERTICAL OVERLAP OF SUBIMAGES
- NHOR NUMBER OF OVERLAPPING HORIZONTAL SUBIMAGES IN A ROW
- IPEND ENDING POINT FOR ONE LINE FROM ERTS
- LASTIM FINAL ROW OF SUBIMAGES
- COR,COV CORRELATION, COVARIANCE MATRICES
- TTL TITLE FOR THE MATRIX
- TITLE TITLE FOR THE MATRIX
- FMT FORMAT FOR PRINTING OUT MATRIX TERMS
- OPT TRUE TO PRINT OUT COVARIANCE MATRIX
- PNCH DETERMINES OUTPUT FILE FOR FEATURES
- TRUE FOR FILE 43, PUNCHED CARDS
- FALSE FOR FILE 01, TAPE OR DISC
- OUTPUT FILE FOR FEATURES
- ROW AND COL NAMES FOR MATRIX PRINTOUT
- DET DETERMINANT OF THE CORRELATION MATRIX
- ENTROPY ENTROPY MEASURE
- KT CARD COUNTER

- SPECTR01
- SPECTR02
- SPECTR03
- SPECTR04
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- SPECTR51
- SPECTR52

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2-12-74 18.629

CROSS-BAND TEXTURE ANALYSIS

```
45 READ(11*IP) (IMAGE(LLN,KOL,I9),IB=1,NDIM)
50 CONTINUE
50 CONTINUE
C           GET THE DIFFERENCE IMAGE
C           CALL DIFFER(IMAGE,X,IXDIM,IYDIM,NDIM,IOIST,NUMPPL,NUMLIN)
C           PUT THE COORDINATE INTO HOLLERITH
C           LITERAL FOR MATRIX PRINTING
C           ENCODE (COL,8) NCOL
C           GET THE COVARIANCE MATRIX FOR THIS
C           SUBIMAGE
C           CALL COVAR(ILINF,NDIM,NUMPPL,X,IXDIM,IYDIM,NUMLIN,MDIN,COV)
C           WRITE OUT THE COVARIANCE MATRIX
C           WHOSE COORDINATES ARE (NCOL,NROW)
C           IF(OPT) CALL SFA07F(COV,NDIM,NDIM,NDIM,2,1,2,FMT,TITLE,TTL(1),
1          TTL(11),TTL(13),LABEL,LABEL)
C           GET THE CORRELATION MATRIX
C           CALL CORREL(COV,NDIM,COR)
C           WRITE OUT THE CORREL MATRIX
C           CALL SFA07F(COR,NDIM,NDIM,NDIM,2,1,2,FM,TITLE,TTL(6),TTL(11),
1          TTL(13),LABFL,LABEL)
C           COPY COR TO COV SINCE HEMDET DESTROYS
C           ITS INPUT MATRIX
C           DO 65 I5=1,NDIM
C           DO 65 J5=1,NDIM
65          COV(I5,J5)=COR(I5,J5)
CONTINUE
C           GET THE DETERMINANT FOR THE
C           CORRELATION MATRIX
C           CALL HEMDET(COV,NDIM,DET)
ENTROP=(-1.)*ALOG(DET)
WRITE(6,91) ENTROP
C           SAVE THE CORRELATION COEFFICIENTS AS
C           VECTORS FOR PRINCIPLE COMPONENTS
C           KT=KT+1
WRITE(IF,200) NROW,NCOL,ENTROP,((COR(J,K+1),K=J,KDIM),J=1,KDIM),KT
90 CONTINUE
91 FORMAT(* ENTROPY MEASURE IS *,F15.5)
.100 CONTINUE
200 FORMAT(2I3,1X,E13.6,6F10.6/8F10.6/8F10.6/6F10.6,15X,I5)
STOP
END
7 MEMORY EXPANDED. USE $LIMITS OR CORE= OPTION FOR NEXT RUN
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REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

CGET.M

GET THE IMAGE FROM ERTS

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GETIM047  
GETIM048  
GETIM049  
GETIM050  
GETIM051  
GETIM052

WRITTEN BY RJ BOSLEY

JAN 1974

DESCRIPTION OF PROGRAM

THIS PROGRAM IS INITIALIZED BY CALLING GETIM. ALL FOLLOWING CALLS MUST BE TO GETIT WHICH GETS THE SUBIMAGE FROM THE ERTS DATA TAPE ON FILE CODE 'ES', AND OUTPUTS A ROW OF OVERLAPPING SUBIMAGES TO DISC FILE 11.

INTERNAL PARAMETERS.

LENGTH  
NOSK

LENGTH OF ONE ERTS LINE OF DATA  
NUMBER OF LINES TO SKIP AFTER  
REWINDING ERTS INPUT TAPE  
STARTING LINE IN ERTS DATA FILE  
VERTICAL INCREMENT FOR THE NEXT ROW  
ARRAY CONTAINING CROSS BAND N-TUPLE  
COMPONENTS

LNSTRT  
LNINCR  
F

ENTRY POINT.

CALL GETIM(ILINE,IDIST,NDIM,IRSTRT,IRSTOP,NUMLIN,NUMPPL,  
LAPVER,LAPHOR,NHOR,INCR,IPEND)

INPUT ARGUMENTS.

ILINE  
IDIST

ARRAY WHERE ERTS LINE IS READ INTO  
DISTANCE BETWEEN NEIGHBORING CELLS  
TO FORM COMPONENTS OF EACH RES CELL  
STARTING LINE OF ERTS DATA FILE  
ENDING LINE OF ERTS DATA FILE  
NUMBER OF LINES IN SUBIMAGE  
NUMBER OF COLUMNS IN SUBIMAGE  
NUMBER OF LINES SUBIMAGES OVERLAP  
NUMBER OF COLUMNS SUBIMAGES OVERLAP

NDIM  
IRSTRT  
IRSTOP  
NUMLIN  
NUMPPL  
LAPVER  
LAP HOR

OUTPUT ARGUMENTS.

NHOR

NUMBER OF HORIZONTAL OVERLAPPING  
SUBIMAGES PER ROW  
HORIZONTAL INCREMENT FOR FIRST COL OF  
THE NEXT SUBIMAGE IN THE ROW  
LAST POINT IN ROW

INCR

IPEND

SUBROUTINE GETIM(ILINE,IDIST,NDIM,IRSTRT,IRSTOP,NUMLIN,NUMPPL,  
LAPVER,LAPHOR,NHOR,INCR,IPEND)

1 DIMENSION ILINE(1),F(8)

INITIALIZE ERTS INPUT TAPE

CALL EINIT(LENGTH)

WRITE(6,4) LENGTH

FORMAT(' LENGTH OF ONE ERTS LINE IS ',16)

CHECK THE NO. OF WORDS PER RECORD

IF(LENGTH.LE.3300) GO TO 5

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ORIGINAL PAGE IS POOR

2-12-74 18.736

GET THE IMAGE FROM ERTS

LA

```
1 WRITE(6,1) LENGTH
  FORMAT(' LENGTH OF ERTS LINE EXCEEDS 3300--LENGTH=',I6)
  STOP
C REWIND THE INPUT TAPE
5 CALL EREWIND
C FIND THE STARTING LINE OF FIRST ROW
C AND INCREMENT FOR SUCCESSIVE ROWS
  IRSTR=IRSTRT-1
  LNSTR=IRSTRT
  LNINCR=NUMLIN-LAPVER
  ASSIGN 951 TO JP
  ASSIGN 971 TO KP
C IF THERE IS NO OVERLAP, DO NOT REWIND
  IF (LAPVER.NE.0) GO TO 940
  ASSIGN 950 TO JP
  ASSIGN 970 TO KP
C FIND ENDING HORIZONTAL POINT IN EACH
C LINE OF THE ROW, AND THE NUMBER OF
C OVERLAPPING SUBIMAGES OF NUMPPL POINTS
C THAT WILL FIT ACROSS THE ROW
940 LEN= (LENGTH-32)/4
  NHOR=0
  IPEND=8
  INCR=NUMPPL-LAPHOR
C INCREMENT THE ENDING POINT
1991 IPEND=IPEND+INCR
  NHOR=NHOR+1
  K=IPEND+NUMPPL+IDIST
C TILL NO MORE SUBIMAGES WILL FIT
  IF (K.LE.LEN) GO TO 1991
  IPEND=IPEND-INCR+NUMPPL+IDIST
C SKIP THE FIRST 8 POINTS IN EACH LINE
  IS=32
C FIND THE LAST POINT IN EACH ERTS LINE
  IF=IPEND*4+32-1
  WRITE(6,955) LEN,IPEND,NHOR
955 FORMAT(' TOTAL POINTS PER ROW IS ',I6 /' FINAL POINT IS ',I6 /
1 ' NUMBER OF OVERLAPPING HORIZONTAL SUBIMAGES IN THE ROW IS ',I5)
  SET THE RANDOM DISC FILE FOR A FIXED
  LENGTH OF NDIM WORDS PER RECORD
C CALL RANSIZ(11,NDIM,1)
C SKIP TO THE LINE BEFORE IRSTR
  IF (IRSTRT.EQ.0) RETURN
  CALL ESKIP(IRSTRT)
  RETURN
C ENTRY POINT TO GET A ROW OF SUBIMAGES
  ENTRY GETIT
C GO TO JP, (950,951)
C CHECK FOR FIRST CALL-IF SO, DONT SKIP
951 IF (LNSTR.EQ.IRSTR) GO TO 950
  NOSK=LNSTR
```

GETIM053  
GETIM054  
GETIM055  
GETIM056  
GETIM057  
GETIM058  
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GETIM098  
GETIM099  
GETIM100  
GETIM101  
GETIM102  
GETIM103  
GETIM104



2-12-74 18.736

GET THE IMAGE FROM ERTS

LA

```
      K=NOSK+NUMLIN
C
      IF(K.LE.1*STOP) GO TO 901
      WRITE(6,900) IRSTOP,NOSK
900  FORMAT(' PROCESSING TERMINATED--NEXT ROW WOULD EXTEND PAST LAST LINE ',I6,' LAST LINE COMPLETED WAS ',I6)
      STOP
C
      SKIP OVER NOSK LINES IN ERTS FILE
901  CALL ESKIP(NOSK)
950  CONTINUE
C
      INCREMENT FOR NEW STARTING LINE
      LNSTRT=LNSTRT+LNINCR
C
      READ IN NUMLIN LINES FOR THE NEW ROW
      KT=0
      DO 903 L=1,NUMLIN
      CALL EREAD(ILINE,LN)
C
      CHECK FOR EOF ON ERTS INPUT TAPE
      IF(LN.NE.0) GO TO 902
      WRITE(6,905)
905  FORMAT(' EOF DETECTED ON ERTS INPUT TAPE--PROCESSING TERMINATED')
      STOP
C
      GO THRU THE LINE CELL BY CELL
902  DO 960 IP=IS,IE,4
      KT=KT+1
C
      GET ALL 8 CROSS-BAND COMPONENTS FOR THIS RESOLUTION CELL
      F(1)=ILINE(IP+2)
      F(2)=(ILINE(IP+2))*ILINE(IP+2)
      F(3)=ILINF(IP+3)
      F(4)=(ILINE(IP+3))*ILINE(IP+3)
      F(5)=(ILINE(IP+2))*ILINE(IP+3)
      F(6)=ILINE(IP+4)
      F(7)=(ILINE(IP+4))*ILINF(IP+4)
      F(8)=(ILINE(IP+2))*ILINE(IP+4)
C
      WRITE OUT THE N-TUPLE OF LENGTH NDIM
      WRITE(11*KT) (F(I),I=1,NDIM)
960  CONTINUE
903  CONTINUE
      GO TO KP,(970,971)
C
      IF THERE IS OVERLAPPING, REWIND TAPE
971  CALL EREWIND
970  RETURN
      END
7 MEMORY EXPANDED. USE $LIMITS OR CORE= OPTION FOR NEXT RUN
```

GETIM105  
GETIM106  
GETIM107  
GETIM108  
GETIM109  
GETIM110  
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GETIM112  
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GETIM142  
GETIM143  
GETIM144  
GETIM145  
GETIM146  
GETIM147

12-12-74 18.839

GET THE DIFFERENCE ARRAY

C DIFFER

GET THE DIFFERENCE ARRAY

DIFFER0

C

DIFFER0

C

WRITTEN BY RJ BOSLEY

JAN 1974

DIFFER0

C

DIFFER0

C

DESCRIPTION OF PROGRAM

DIFFER0

C

THIS PROGRAM REPLACES THE ORIGINAL IMAGE IA WITH THE NEAREST

DIFFER0

C

NEIGHBOR HORIZONTAL DIFFERENCE, (I1-J1, I2-J2, ..., IN-JN) WHERE

DIFFER0

C

I AND J ARE N-DIMENSIONAL HORIZONTALLY NEIGHBORING RESOLUTION

DIFFER0

C

CELLS OF DISTANCE IDIST. NOTE THAT THE ABSOLUTE VALUE IS USED

DIFFER0

C

GIVING ONLY THE POSITIVE HALF OF THE DISTRIBUTION OF

DIFFER1

C

DIFFERENCES I-J AND J-I.

DIFFER1

C

ENTRY POINT.

DIFFER1

C

CALL DIFFER(IA,X,IXDIM,IYDIM,NDIM,IDIST,NUMPPL,NUMLIN)

DIFFER1

C

ARGUMENTS.

DIFFER1

C

IA

THE SUBIMAGE BEING PROCESSED

DIFFER1

C

X

DIFFERENCE ARRAY

DIFFER1

C

IXDIM

COL DIMENSION OF X

DIFFER1

C

IYDIM

ROW DIMENSION OF X

DIFFER1

C

NDIM

THE DIMENSION OF EACH RESOLUTION CELL

DIFFER2

C

IDIST

DISTANCE BETWEEN RESOLUTION CELLS

DIFFER2

C

NUMPPL

USED FOR THE DIFFERENCE ARRAY

DIFFER2

C

NUMLIN

NUMBER OF COLUMNS IN SUBIMAGE

DIFFER2

C

NUMBER OF LINES IN SUBIMAGE

DIFFER2

C

SUBROUTINE DIFFER(IA,X,IXDIM,IYDIM,NDIM,IDIST,NUMPPL,NUMLIN)

DIFFER2

C

DIMENSION IA(IYDIM,IXDIM,1), X(IYDIM,IXDIM,1)

DIFFER2

C

GO THRU EACH DIMENSION OR BAND

DIFFER2

C

DO 5 IBAND=1,NDIM

GO THRU EACH LINE

DIFFER2

C

DO 4 LINE=1,NUMLIN

GO THRU ALL BUT THE LAST COLUMN

DIFFER2

C

DO 3 KOL=1,NUMPPL

REPLACE EACH RESOLUTION CELL COMPONENT

DIFFER3

C

KKOL=KOL+IDIST

BY THE DIFFERENCE

DIFFER3

C

X(LINE,KOL,IBAND)=IABS(IA(LINE,KOL,IBAND)-IA(LINE,KKOL,IBAND))

DIFFER3

C

CONTINUE

DIFFER3

3

CONTINUE

DIFFER3

4

CONTINUE

DIFFER3

5

RETURN

DIFFER3

END

DIFFER3

DIFFER3

DIFFER3

DIFFER4

DIFFER4

DIFFER4

02-12-74 18.927

FIND THE COVARIANCE MATRIX FOR THE SUBIMAGE

```
CCOVAR          FIND THE COVARIANCE MATRIX FOR THE SUBIMAGE          COVAR001
C                                                       COVAR002
C                                                       COVAR003
C   WRITTEN BY RJ BOSLEY          JAN 1974          COVAR004
C                                                       COVAR005
C   DESCRIPTION OF PROGRAM          COVAR006
C   THIS PROGRAM TAKES THE DIFFERENCE SUBIMAGE AND CALCULATES THE          COVAR007
C   COVARIANCE MATRIX FOR IT.          COVAR008
C                                                       COVAR009
C   DESCRIPTION OF PARAMETERS          COVAR010
C   PERCENT          PERCENTAGE OF TOTAL VECTORS X FROM          COVAR011
C   SCR,XLINE,SAM          WHICH COVARIANCE IS CALCULATED          COVAR012
C   IER          SCRATCH ARRAYS          COVAR013
C   JER          ERROR FLAG FROM MNCVIN          COVAR014
C   ENTRY POINT.          ERROR FLAG FROM MNCV          COVAR015
C   CALL COVAR(XLINE,NDIM,NUMPPL,X,IXDIM,IYDIM,NUMLIN,NDIN,COV)          COVAR016
C   INPUT ARGUMENTS.          COVAR017
C   XLINE          ARRAY USED TO SEND ONE LINE TO MNCV          COVAR018
C   NDIM          DIMENSION OF VECTORS X, NO OF BANDS          COVAR019
C   NUMPPL          NUMBER OF COLUMNS IN SUBIMAGE          COVAR020
C   X          FLOATING POINT DIFFERENCE ARRAY          COVAR021
C   IXDIM,IYDIM          COL,ROW DIMENSIONS OF X          COVAR022
C   NUMLIN          NUMBER OF LINES IN SUBIMAGE          COVAR023
C   NDIN          DIMENSION OF COV ARRAY          COVAR024
C                                                       COVAR025
C   OUTPUT ARGUMENTS.          COVAR026
C   COV          COVARIANCE MATRIX FOR THE SUBIMAGE          COVAR027
C                                                       COVAR028
C   SUBROUTINE COVAR(XLINE,NDIM,NUMPPL,X,IXDIM,IYDIM,NUMLIN,NDIN,COV)          COVAR029
C   DIMENSION X(IYDIM,IXDIM,1),XLINE(NDIM,NUMPPL),XMEAN(8),SCR(8),          COVAR030
C   1 NTRUTH(1),SAM(1),COV(NDIN,NDIN)          COVAR031
C   PERCENT=100.0          COVAR032
C   CALL MNCVIN(NUMPPL,NDIM,NUMLIN,PERCENT,1,XLINE,NTRUTH,COV,XMEAN,          COVAR033
C   1 SCR,SAM,IER,JER)          COVAR034
C   INITIALIZE THE COVARIANCE PROGRAM          COVAR035
C   CHECK FOR AN ERROR          COVAR036
C   IF(IER.EQ.0) GO TO 1          COVAR037
C   WRITE(6,2) IER          COVAR038
C   2 FORMAT(' ERROR IN MNCVIN, IERROR IS ',I3)          COVAR039
C   STOP          COVAR040
C   1 CONTINUE          COVAR041
C   GO THRU EACH LINE OF SUBIMAGES          COVAR042
C   DO 10 LINE=1,NUMLIN          COVAR043
C                                                       COVAR044
```

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FIND THE COVARIANCE MATRIX FOR THE SUBIMAGE

```
C          GET ONE LINE OF DATA          COVAR046
          DO 6 I=1,NDIM                   COVAR047
          DO 5 KOL=1,NUMPPL                COVAR048
5         X_LINE(I,KOL)=X(LINE,KOL,I)     COVAR049
6         CONTINUE                        COVAR050
          INCREMENT COVARIANCE CALCULATIONS COVAR051
          CALL MNCV                         COVAR052
          CHECK FOR GROUND TRUTH ERROR     COVAR053
          IF (JER.EQ.0) GO TO 10           COVAR054
          WRITE(6,7) JER                   COVAR055
7         FORMAT(' ERROR IN MNCV, JERROR IS ',I2) COVAR056
          STOP                              COVAR057
10        CONTINUE                         COVAR058
          NOTE***ONLY THE POSITIVE DIFFERENCES COVAR059
          WERE USED IN THE CALCULATIONS---THE COVAR060
          TRUE MEAN MUST BE ZERO. SO WE MUST COVAR061
          ADD XMEAN**2 TO EACH ELEMENT     COVAR062
          DO 20 I=1,NDIM                   COVAR063
          DO 20 J=I,NDIM                   COVAR064
          COV(I,J)=COV(I,J)+XMEAN(I)*XMEAN(J) COVAR065
20        COV(J,I)=COV(I,J)              COVAR066
          RETURN                            COVAR067
          END                               COVAR068
```

CMNCV

M N C V I N - M N C V

IDENTIFICATION

PROGRAM NAME	MNCVIN
OTHER ENTRY POINT	MNCV
SYSTEM	POP-15
SOURCE LANGUAGE	FORTRAN IV
AUTHOR	JAMES D YOUNG
DATE	8/18/73

PURPOSE

TO CALCULATE THE MEAN VECTOR AND COVARIANCE MATRIX FOR EACH CATEGORY OF A SET OF VECTORS. THE CALCULATIONS WILL BE BASED ON A SPECIFIED PERCENTAGE OF THE VECTORS RANDOMLY CHOSEN WITHIN THE SET.

ENTRY POINT - MNCVIN(NVPCAL,NDIM,NCALL,PERCNT,NCAT,X,NTRUTH,COV,XMEAN,SCTMEN,SAMSZ,IERROR,JERROR)

THIS INITIALIZES THE ROUTINE. AFTER MNCVIN HAS BEEN CALLED CHECK IERROR TO SEE IF IT IS NONZERO WHICH INDICATES THAT AN ERROR HAS OCCURRED.

ENTRY POINT - MNCV

THE CALL TO MNCV SHOULD BE PERFORMED NCALL TIMES, EACH TIME WITH THE NEXT GROUP OF VECTORS IN X. IF MORE THAN ONE CATEGORY IS BEING CONSIDERED, THE GROUND TRUTH INTEGERS IN NTRUTH ASSOCIATED WITH THE VECTORS SHOULD ALSO BE UPDATED EACH TIME MNCV IS CALLED, AND JERROR SHOULD BE CHECKED AFTER EACH CALL TO SEE IF IT IS NONZERO WHICH INDICATES THAT AN ILLEGAL GROUND TRUTH INTEGER HAS BEEN FOUND. IF ONLY ONE CATEGORY IS BEING CONSIDERED HOWEVER, THE VALUES IN NTRUTH ARE NOT USED IN THE ROUTINE.

AFTER MNCV HAS BEEN CALLED NCALL TIMES THE MEAN VECTOR AND COVARIANCE MATRIX FOR EACH CATEGORY IS COMPLETED. IN ADDITION, THE NUMBER OF VECTORS USED FOR THE CALCULATIONS FOR EACH CATEGORY IS OUTPUT. THESE NUMBERS SHOULD BE CHECKED WHEN APPROPRIATE TO SEE IF ENOUGH VECTORS WERE USED FROM EACH CATEGORY TO GIVE A REASONABLE ESTIMATE OF THE DESIRED STATISTICS.

ARGUMENTS

INPUT -

NVPCAL NUMBER OF VECTORS PER CALL

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NDIM DIMENSION OF DATA VECTORS  
 NCALL NUMBER OF CALLS  
 PERCNT PERCENTAGE OF TOTAL NUMBER OF VECTORS FROM WHICH  
 MEAN AND COVARIANCE MATRICES WILL BE CALCULATED  
 NCAT NUMBER OF CATEGORIES CONSIDERED  
 =1 IF ONLY ONE SET OF STATISTICS WILL BE  
 CALCULATED FOR ALL DATA  
 =NUMBER OF CATEGORIES IN DATA IF ONE SET OF  
 STATISTICS WILL BE CALCULATED FOR EACH CATEGORY  
 X X(NDIM,NVPCAL) MATRIX CONTAINING INPUT DATA VECTORS  
 IN ITS COLUMNS  
 NTRUTH NTRUTH(NVPCAL) VECTOR CONTAINING THE GROUND TRUTH  
 INTEGERS, 1 THROUGH NCAT, ASSOCIATED WITH THE DATA  
 VECTORS OF X. IF NCAT EQUALS 1 THIS VECTOR WILL NOT  
 BE USED.

## OUTPUT -

COV COV(NDIM,NDIM,NCAT) MATRIX CONTAINING COVARIANCE  
 MATRIX/MATRICES OF THE DATA  
 XMEAN XMEAN(NDIM,NCAT) MATRIX CONTAINING MEAN  
 VECTOR/VECTORS OF THE DATA  
 SCTMEN SCTMEN(NDIM,NCAT) SCRATCH MATRIX CONTAINING AN  
 ESTIMATE OF THE MEAN VECTOR/VECTORS  
 SAMSZ SAMSZ(NCAT) VECTOR CONTAINING NUMBER OF VECTORS USED  
 TO CALCULATE THE STATISTICS FOR EACH CATEGORY  
 IERROR ERROR INDICATED IF RETURNED NONZERO  
 =1 IF NVPCAL .LE. 0  
 =2 IF NDIM .LE. 0  
 =3 IF NCAL .LE. 0  
 =4 IF PERCNT .GT. 100. OR PERCNT IS SO SMALL THAT  
 FEWER THAN 2 VECTORS WILL BE USED TO CALCULATE  
 ALL THE STATISTICS  
 =5 IF NCAT .LE. 0  
 (IF MORE THAN ONE OF THESE ERRORS OCCURS,  
 THE HIGHER VALUE WILL BE RETURNED)

JERROR ERROR INDICATED IF RETURNED NONZERO  
 =1 IF ILLEGAL GROUND TRUTH LABEL IS FOUND

INTERNAL PARAMETERS

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02-09-74 20.440

M N C V I N - M N C V

C ICALL NUMBER OF TIMES MNCV HAS BEEN CALLED  
C VECTS TOTAL NUMBER OF VECTORS WHICH WILL BE INPUT TO MNCV  
C VTBU TOTAL NUMBER OF VECTORS TO BE USED IN CALCULATION  
C OF STATISTICS  
C VLTBU NUMBER OF VECTORS LEFT TO BE USED IN CALCULATION  
C OF STATISTICS  
C VLEFT NUMBER OF VECTORS LEFT TO BE CONSIDERED  
C INTRU INTEGER DENOTING GROUND TRUTH CATEGORY  
C IP ARGUMENT TO RCM  
C IQ ARGUMENT TO RCM

-----  
INITIALIZER ENTRY POINT

C SUBROUTINE MNCVIN(NVPCAL,NDIM,NCALL,PERCNT,NCAT,X,NTRUTH,COV,  
C 1 XMEAN,SCTMEN,SAMSZ,IERROR,JERROR)

C DIMENSION X(NDIM,1),NTRUTH(1),COV(NDIM,NDIM,NCAT),XMEAN(NCAT,1),  
C 1 SCTMEN(NDIM,1),SAMSZ(NCAT)

C ICALL=0  
C VECTS=NVPCAL\*NCALL  
C IERROR=0  
C NVTBU=VECTS\*PERCNT/100.+0.49999  
C VTBU=NVTBU  
C VLTBU=VTBU  
C VLEFT=VECTS  
C INTRU=1  
C IP=33333  
C IQ=55555  
C JERROR=0

C CHECK LEGALITY OF SOME NUMBERS :

C IF(NVPCAL.LE.0) IERROR=1  
C IF(NDIM .LE.0) IERROR=2  
C IF(NCALL .LE.0) IERROR=3  
C IF(PERCNT.GT.100..OR.VTBU.LT.2.) IERROR=4  
C IF(NCAT .LE.0) IERROR=5

C ZERO OUT A FEW ARRAYS

C DO 14 K=1,NCAT  
C SAMSZ(K)=0  
C DO 14 J=1,NDIM  
C XMEAN(J,K)=0  
C SCTMEN(J,K)=0  
C DO 14 I=1,NDIM  
C 14 COV(I,J,K)=0

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR

02-09-74 20.440

M N C V I N - M N C V

SET TRANSFER TO REFLECT NUMBER  
OF CATEGORIES

C  
C  
C

ASSIGN 5 TO IGO  
IF(NCAT.GT.1) ASSIGN 15 TO IGO  
RETURN

C  
C  
C  
C  
C

\*\*\*\*\*  
ENTRY POINT FOR ALL CALLS AFTER THE FIRST

ENTRY MNCV

ICALL=ICALL+1  
IF((ICALL/50)\*50.NE.ICALL.AND.ICALL.NE.10) GO TO 1

C  
C  
C  
C  
C

UPDATE ESTIMATE OF THE MEANS AND MODIFY  
COVARIANCE CALCULATIONS TO REFLECT  
THIS UPDATE

DO 10 K=1,NCAT  
IF(SAMZ(K).EQ.0.) GO TO 10  
DO 2 J=1,NDIM  
DO 2 I=1,J  
2 COV(I,J,K)=COV(I,J,K)-SAMZ(K)\*(XMEAN(I,K)/SAMZ(K)-SCTMEN(I,K))\*  
1 (XMEAN(J,K)/SAMZ(K)-SCTMEN(J,K))  
DO 3 J=1,NDIM  
SCTMEN(J,K)=XMEAN(J,K)/SAMZ(K)  
3 CONTINUE  
10 CONTINUE

C  
C  
C  
C

CONTINUE TO CALCULATE MEAN AND  
UPPER TRIANGLE OF COVARIANCE MATRICES

1 DO 8 I=1,NVPCAL

C  
C  
C  
C

DETERMINE WHETHER TO SKIP THIS VECTOR

IF(RCM(IP,IQ).GT.VLTBU/VLEFT) GO TO 11

C  
C  
C  
C

INCLUDE THIS VECTOR IN CALCULATIONS

VLTBU=VLTBU-1.  
GO TO IGO,(15,5)

C  
C  
C  
C

WE ARE TO CONSIDER MORE THAN ONE  
CATEGORY

15 INTRU=NTRUTH(I)

C  
C  
C  
C

CHECK LEGALITY OF GROUND TRUTH LABEL

REPRODUCIBILITY OF THE  
ORIGINAL PAGE IS POOR



02-09-74 20.440

M N C V I N - M N C V

IF (INTRU.LE.0.OR.INTRU.GT.NCAT) GO TO 9  
5 DO 4 J=1,NOIM

C  
C  
C

SUM FOR MEAN

XMEAN(J,INTRU)=YMEAN(J,INTRU)+X(J,I)  
DO 4 K=1,J

C  
C  
C

SUM FOR FOR COVARIANCE

COV(K,J,INTRU)=COV(K,J,INTRU)+(X(J,I)-SCTMEN(J,INTRU))\*(X(K,I)-  
1 SCTMEN(K,INTRU))  
4 CONTINUE  
SAMSZ(INTRU)=SAMSZ(INTRU)+1.  
11 CONTINUE  
VLEFT=VLEFT-1.  
8 CONTINUE  
IF(ICALL.LT.NCALL) RETURN

C  
C  
C  
C

ALL VECTORS HAVE BEEN INPUT  
FINISH CALCULATION OF STATISTICS

DO 7 K=1,NCAT  
IF(SAMSZ(K).EQ.0.) GO TO 7  
DO 6 J=1,NOIM  
DO 6 I=1,J

COV(I,J,K)=COV(I,J,K)/SAMSZ(K)-(XMEAN(I,K)/SAMSZ(K)-SCTMEN(I,K))\*  
1 (XMEAN(J,K)/SAMSZ(K)-SCTMEN(J,K))

C  
C  
C

FILL IN LOWER TRIANGLE

6 COV(J,I,K)=COV(I,J,K)  
DO 13 J=1,NOIM  
XMEAN(J,K)=XMEAN(J,K)/SAMSZ(K)  
13 CONTINUE  
7 CONTINUE  
12 CONTINUE  
RETURN  
9 JERROR=1  
GO TO 12  
END

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ORIGINAL PAGE IS POOR

02-12-74 18.936

GET THE CORRELATION MATRIX FOR THE SUBIMAGE

LA

CCORPEL

GET THE CORRELATION MATRIX FOR THE SUBIMAGE

CORREL01  
 CORREL02  
 CORREL03  
 CORREL04  
 CORREL05  
 CORREL06  
 CORREL07  
 CORREL08  
 CORREL09  
 CORREL10  
 CORREL11  
 CORREL12  
 CORREL13  
 CORREL14  
 CORREL15  
 CORREL16  
 CORREL17  
 CORREL18  
 CORREL19  
 CORREL20  
 CORREL21  
 CORREL22  
 CORREL23  
 CORREL24  
 CORREL25  
 CORREL26  
 CORREL27  
 CORREL28

JAN 1974

WRITTEN BY RJ BOSLEY

DESCRIPTION OF PROGRAM

THIS PROGRAM CALCULATES THE CORRELATION MATRIX GIVEN THE  
COVARIANCE MATRIX AND ITS ORDER.

ENTRY POINT.

CALL CORREL(COV,NDIM,COR)  
ARGUMENTS.

COV	COVARIANCE MATRIX ARRAY
NDIM	ORDER OF MATRICES
COR	CORRELATION MATRIX ARRAY

SUBROUTINE CORREL(COV,NDIM,COR)

DIMENSION COV(NDIM,NDIM),COR(NDIM,NDIM)  
GO THRU EACH ROW OR LINE

DO 10 LINE=1,NDIM  
 COVL=ABS(COV(LINE,LINE))  
 GO THRU UPPER DIAGONAL

DO 9 KOL=LINE,NDIM  
 COVC=ABS(COV(KOL,KOL))  
 COR(LINE,KOL)=COV(LINE,KOL)/SQRT(COVC\*COVL)  
 COR(KOL,LINE)=COR(LINE,KOL)  
 CONTINUE  
 RETURN  
 END

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ORIGINAL PAGE IS POOR

IV.2-d Piecewise Linear Classification Program Listings

RCLASS  
XIN  
LINEAR  
WEIGHT

CRCLASS

## PATTERN VECTOR CLASSIFICATION

WRITTEN BY SAM SHANMUGAM  
DOCUMENTED BY RJ BOSLEY

SEPT 1972  
DEC 1973

RCLASS01  
RCLASS02  
RCLASS03  
RCLASS04  
RCLASS05  
RCLASS06  
RCLASS07  
RCLASS08  
RCLASS09  
RCLASS10  
RCLASS11  
RCLASS12  
RCLASS13  
RCLASS14  
RCLASS15  
RCLASS16  
RCLASS17  
RCLASS18  
RCLASS19  
RCLASS20  
RCLASS21  
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RCLASS32  
RCLASS33  
RCLASS34  
RCLASS35  
RCLASS36  
RCLASS37  
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RCLASS39  
RCLASS40  
RCLASS41  
RCLASS42  
RCLASS43  
RCLASS44  
RCLASS45  
RCLASS46  
RCLASS47  
RCLASS48  
RCLASS49  
RCLASS50

## DESCRIPTION OF PROGRAM.

THIS SUBROUTINE CLASSIFIES A GIVEN SET OF PATTERN VECTORS INTO ONE OF MANY POSSIBLE CATEGORIES USING THE INFORMATION CONTAINED IN A SET OF TRAINING PATTERNS WHOSE CATEGORIES ARE KNOWN.

## ENTRY POINT.

CALL RCLASS(WORK,ISIZE)

## INPUT ARGUMENTS.

WORK - SCRATCH ARRAY USED BY RCLASS

ISIZE - SIZE OF THE SCRATCH ARRAY WORK. THE ARRAY WORK MUST BE DIMENSIONED IN THE CALLING PROGRAM, WITH A DIMENSION GREATER THAN OR EQUAL TO ISIZE. THE VALUE OF ISIZE IS GIVEN BY

$$ISIZE = NDIM * (NTRAIN + 10) + 1000 \\ + (NC * (NC + 1) / 2) * ND + NPAIR * ND$$

WHERE  $ND = NDIM - 1$

## INPUT PARAMETER CARDS.

CARD 1. SHOULD CONTAIN THE PROGRAM OPTION PARAMETERS NOPT1, NOPT2, NOPT3, NOPT4, IN FORMAT(4I1).

IF(NOPT1.NE.0) THE TRAINING PATTERNS ARE PRINTED OUT  
IF(NOPT2.NE.0) THE TEST PATTERNS ARE PRINTED OUT  
IF(NOPT3.NE.0) THE CLASSIFICATION OF EACH TRAINING PATTERN IS LISTED. OTHERWISE, ONLY THE CONTINGENCY TABLE IS PRINTED OUT  
IF(NOPT4.NE.0) THE CLASSIFICATION OF EACH TEST PATTERN IS LISTED. OTHERWISE ONLY THE CONTINGENCY TABLE IS PRINTED OUT.

CARD 2. SHOULD CONTAIN PARAMETERS NTOT, NPART, NDIM, NC, NPAIR IN FORMAT(5I5).

NTOT -- TOTAL NUMBER OF PATTERN VECTORS IN THE DATA SET  
NPART -- NPART OUT OF EVERY 10 PATTERN VECTORS IN THE DATA SET WILL BE USED FOR TRAINING. THE REMAINING PATTERNS WILL BE USED FOR TESTING THE CLASSIFIER  
NDIM -- NUMBER OF MEASUREMENTS PER VECTOR PLUS 2  
NC -- NUMBER OF GROUND TRUTH CATEGORIES  
NPAIR -- 2\*(MAXIMUM NUMBER OF TRAINING PATTERNS IN

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## ANY ONE CATEGORY)

## INPUT DATA SET.

INPUT DATA VECTORS ARE READ IN BINARY FROM FILE CODE 01  
ACCORDING TO THE FOLLOWING READ---

WHERE IGT READ(01) IGT,M1,N1,NFT,(FEAT(I),I=1,NMEAS)  
IS THE GROUND TRUTH CATEGORY  
M1,N1,NFT ARE NOT USED (MAY BE USED AS TAGS)  
FEAT(I) IS THE FEATURE VECTOR I-TH COMPONENT  
NMEAS IS THE NUMBER OF COMPONENTS OR MEASUREMENTS IN  
EACH FEATURE VECTOR=NDIM-2.

THE DATA FILE SHOULD HAVE A TOTAL OF NTOT LOGICAL RECORDS IN  
BINARY FORM. EACH LOGICAL RECORD MUST BE OF LENGTH (NDIM+2)  
WORDS. WORD 1 IS THE GROUND TRUTH CATEGORY AND 5 THRU NDIM+2 ARE  
THE MEASUREMENT VALUES. THE PATTERN VECTORS MUST BE SORTED BY  
CATEGORY. THE DATA SET IS SORTED INTO TRAINING AND TEST SETS  
ACCORDING TO THE USER SPECIFIED RATIO IN NPART. TRAINING PATTERNS  
ARE COPIED TO CORE AND TEST PATTERNS ARE COPIED TO DISC ON FILE  
CODE 02.

## LIMITATIONS.

1. MAXIMUM NUMBER OF CATEGORIES IS 15.
2. MAXIMUM VALUE OF NDIM IS 100.
3. SCRATCH DISC FILE MUST BE ON 02 AND INPUT DATA ON 01.
4. PATTERN VECTORS MUST BE SORTED BY CATEGORY.

## THEORY.

USING A REGRESSION TYPE ALGORITHM, THE PROGRAM  
OBTAINS A SET OF HYPERPLANES FOR SEPARATING THE  
TRAINING PATTERNS. THE SEPARATION IS DONE FOR  
DIFFERENT CATEGORY PAIRS. A TOTAL OF  $NC*(NC-1)/2$   
HYPER PLANES ARE DETERMINED. TEST PATTERNS ARE  
CLASSIFIED ON THE BASIS OF A MAJORITY VOTE ON THESE  
HYPERPLANES.

## BIBLIOGRAPHY.

FOR A COMPLETE DISCUSSION OF DETAILS, SEE "INTRODUCTION TO  
STATISTICAL PATTERN RECOGNITION" BY FUKUNAGA, ACADEMIC PRESS,  
NEW YORK, 1972.

## SUBPROGRAMS REQUIRED.

DRIVER

RCLASS

XIN

LINEAR

WEIGHT

MINV -- A FORTRAN CALLABLE ROUTINE FROM IBM SSP

## EXAMPLE.

IN ORDER TO PROCESS A DATA SET REQUIRING A SCRATCH ARRAY OF

RCLASS51  
RCLASS52  
RCLASS53  
RCLASS54  
RCLASS55  
RCLASS56  
RCLASS57  
RCLASS58  
RCLASS59  
RCLASS60  
RCLASS61  
RCLASS62  
RCLASS63  
RCLASS64  
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RCLASS66  
RCLASS67  
RCLASS68  
RCLASS69  
RCLASS70  
RCLASS71  
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RCLASS82  
RCLASS83  
RCLASS84  
RCLASS85  
RCLASS86  
RCLASS87  
RCLASS88  
RCLASS89  
RCLASS90  
RCLASS91  
RCLASS92  
RCLASS93  
RCLASS94  
RCLASS95  
RCLASS96  
RCLASS97  
RCLASS98  
RCLASS99  
RCLASS00



CXIN	READ IN THE DATA	XIN00001
C		XIN00002
C	WRITTEN BY SAM SHANMUGAM	SEPT 1972 XIN00003
C	DOCUMENTED BY RJ BOSLEY	DEC 1973 XIN00004
C		XIN00005
C	DESCRIPTION OF PROGRAM.	XIN00006
C	THIS SUBROUTINE READS IN THE TWO PARAMETER CARDS AND THEN	XIN00007
C	READS IN THE DATA SET, COPYING THE TEST PATTERNS TO DISC FILE 02.	XIN00008
C		XIN00009
C	ENTRY POINT.	XIN00010
C	CALL XIN(WORK,U)	XIN00011
C		XIN00012
C	INPUT ARGUMENTS.	XIN00013
C	WORK	XIN00014
C	U	XIN00015
C		XIN00016
C		XIN00017
C	INTERNAL PARAMETERS.	XIN00018
C	NTRAIN	XIN00019
C	NTEST	XIN00020
C	NDIM	XIN00021
C	NO	XIN00022
C	NC	XIN00023
C	NPAIR	XIN00024
C	NH	XIN00025
C	NOPT1-4	XIN00026
C	NSKIP	XIN00027
C		XIN00028
C	NTIMES	XIN00029
C		XIN00030
C	ILEFT	XIN00031
C		XIN00032
C		XIN00033
C		XIN00034
C	SUBROUTINE XIN(WORK,U)	XIN00035
C		XIN00036
C	DIMENSION WORK(1),U(1)	XIN00037
C	COMMON NTRAIN,NTEST,NOIM,NO,NC,NPAIR,NH,NOPT1,NOPT2,NOPT3,NOPT4,	XIN00038
C	1 NTOT,NPART	XIN00039
C		XIN00040
C	READ(5,100) NOPT1,NOPT2,NOPT3,NOPT4	XIN00041
C		XIN00042
C	READ(5,101) NTOT,NPART,NOIM,NC,NPAIR	XIN00043
C		XIN00044
100	FORMAT(4I1)	XIN00045
101	FORMAT(5I5)	XIN00046
C		XIN00047
C		XIN00048
C		XIN00049
C	REWIND 01	XIN00050
C		
C	I=REGIN=1	
	ARRAY DATA IS READ INTO SCRATCH ARRAY	
	NO. OF TRAINING PATTERN VECTORS	
	NO. OF TEST PATTERN VECTORS	
	2+NO. OF MEAS PER FEATURE VECTOR	
	NOIM-1	
	NO. OF GROUND TRUTH CATEGORIES	
	TWICE MAX NO. OF TRAINING PATTERNS	
	NO. OF HYPERPLANES	
	PROGRAM OPTIONS, SEE RCLASS	
	NO. OF VECTORS OUT OF 10 TO SKIP TRAINING	
	NO. OF BLOCKS OF 10 PATTERN VECTORS IN INPUT DATA FILE	
	NO. OF VECTORS LEFT AFTER READING ALL BLOCKS OF 10 VECTORS	

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	NTRAIN=0			XIN00051
	NDD=NDIM+2			XIN00052
	NTIMES=NTOT/10			XIN00053
C			OUT OF EACH OF THE NTIMES BLOCKS OF	XIN00054
C			TEN INPUT PATTERNS PICK THE FIRST	XIN00055
C			NPART PATTERNS AS TRAINING PATTERNS	XIN00056
	NSKIP=10-NPART			XIN00057
C				XIN00058
C			WRITE HEADING	XIN00059
	IF (NOPT1.NE.0) WRITE(6,201)			XIN00060
201	FORMAT(1H1,10X,'TRAINING PATTERNS')			XIN00061
C			GO THRU EACH BLOCK OF TEN	XIN00062
C			PATTERN VECTORS	XIN00063
	DO 20 NN=1,NTIMES			XIN00064
	IF (NSKIP.EQ.0) GO TO 16			XIN00065
C			SKIP OVER THE TEST VECTORS	XIN00066
	DO 15 JJ=1,NSKIP			XIN00067
C			READ PAST THIS VECTOR	XIN00068
15	READ(1) (U(K),K=1,NDD)			XIN00069
C				XIN00070
16	CONTINUE			XIN00071
C			READ NPART TRAINING PATTERN VECTORS	XIN00072
	DO 10 JJ=1,NPART			XIN00073
C			COUNT THE NO. OF TRAINING VECTORS	XIN00074
	NTRAIN=NTRAIN+1			XIN00075
C			INDEX THIS VECTOR INTO WORK	XIN00076
	IBEGIN=(NTRAIN-1)*NDIM+1			XIN00077
	IEND=NTRAIN*NDIM			XIN00078
	IB1=IBEGIN+1			XIN00079
	IB2=IB1+1			XIN00080
C			SET THE FIRST COMPONENT TO 1.0	XIN00081
	WORK(IB1)=1.0			XIN00082
C			READ THE TRAINING PATTERN VECTORS	XIN00083
	READ(1) WORK(IBEGIN),NM1,NM2,NM3,(WORK(K),K=IB2,IEND)			XIN00084
C			WRITE OUT THE TRAINING PATTERNS	XIN00085
	IF (NOPT1.NE.0) WRITE(6,102) NTRAIN,WORK(IBEGIN),NM1,NM2,(WORK(K),			XIN00086
102	1K=IB1,IEND)			XIN00087
10	FORMAT(1X,I3,3X,6(F10.4))			XIN00088
20	CONTINUE			XIN00089
	CONTINUE			XIN00090
C			READ IN THE LEFTOVER PATTERNS	XIN00091
	ILEFT=NTOT-10*NTIMES			XIN00092
	IF (ILEFT.EQ.0) GO TO 26			XIN00093
	DO 25 JJ=1,ILEFT			XIN00094
C			INCREMENT TRAINING VECTOR COUNT	XIN00095
	NTRAIN=NTRAIN+1			XIN00096
C			INDEX THE VECTOR INTO WORK	XIN00097
	IBEGIN=(NTRAIN-1)*NDIM+1			XIN00098
	IEND=NTRAIN*NDIM			XIN00099
	IB1=IBEGIN+1			XIN00100



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READ IN THE DATA

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I92=I91+1
WORK(IB1)=1.0
C READ(1) WORK(IBEGIN),NM1,NM2,NM3,(WORK(K),K=I92,IEND) READ A TRAINING PATTERN
C IF(NOPT1.NE.0) WRITE(6,102) NTRAIN,WORK(IBEGIN),NM1,NM2,(WORK(K), WRITE OUT TRAINING PATTERNS
1K=I91,IEND)
C
25 CONTINUE
26 CONTINUE
REWIND 01
REWIND 02
C COPY TEST PATTERNS TO DISC FILE 02
C WRITE HEADING
C IF(NOPT2.NE.0) WRITE(6,103)
103 FORMAT(1H1,10X,'TEST PATTERNS')
NTEST=0
C SKIP IF THERE ARE NO TEST PATTERNS
C IF(NPART.EQ.0) GO TO 99
DO 45 NN=1,NTIMES
DO 43 JJ=1,NSKIP
C COUNT THE NO. OF TEST VECTORS
C NTEST=NTEST+1
C READ INPUT DATA FILE THE SECOND TIME
C READ(1) (U(KK),KK=1,NDD)
C SET COMPONENT FOUR TO 1.0
C U(4)=1.0
C WRITE OUT THE TEST PATTERNS
C IF(NOPT2.NE.0) WRITE(6,102) NTEST,U(1),U(2),U(3),(U(KK),KK=4,NDD)
C WRITE TEST PATTERN TO DISC FILE 02
43 WRITE(2),(U(KK),KK=1,NDD)
C SKIP OVER THE TRAINING VECTORS
C DO 44 JJ=1,NPART
44 READ(1) (U(KK),KK=1,NDD)
45 CONTINUE
END FILE 02
REWIND 02
99 CONTINUE
REWIND 01
C WRITE OUT A PROGRESS NOTE
702 WRITE(6,702)
FORMAT(5X,'DATA HAS BEEN COPIED ON TO DISK 02')
C DETERMINE THE NO. OF HYPERPLANES
NH=NC*(NC+1)/2
ND=NDIM-1
RETURN
END
XIN00101
XIN00102
XIN00103
XIN00104
XIN00105
XIN00106
XIN00107
XIN00108
XIN00109
XIN00110
XIN00111
XIN00112
XIN00113
XIN00114
XIN00115
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XIN00146
XIN00147
XIN00148
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CLINEAR          LINEAR DISCRIMINANT FUNCTION          LINEAR01
C                                                       LINEAR02
C                                                       LINEAR03
C   WRITTEN BY SAM SHANMUGAM          SEPT 1972        LINEAR04
C   DOCUMENTED BY RJ BOSLEY          DEC 1973          LINEAR05
C                                                       LINEAR06
C   DESCRIPTION OF PROGRAM.          LINEAR07
C   THIS SUBROUTINE CALLS THE REGRESSION ROUTINE WEIGHT TO GET THE LINEAR08
C   DISCRIMINATION FUNCTIONS, COMBINES THESE TO OBTAIN A PIECEWISE LINEAR09
C   LINEAR DISCRIMINANT FUNCTION FOR EACH CATEGORY, AND CLASSIFIES LINEAR10
C   THE TRAINING AND TEST PATTERNS. THIS ROUTINE ALSO OUTPUTS THE LINEAR11
C   RESULTS.                          LINEAR12
C                                                       LINEAR13
C   THEORY.                            LINEAR14
C   THE ALGORITHM FOR DETERMINING THE WEIGHT VECTOR FOR SEPARATING LINEAR15
C   THE I AND J-TH CATEGORIES IS  $w = ((U*U)^{-1}) * (U*T)$ , WHERE T IS LINEAR16
C   A VECTOR OF LENGTH ND WITH COMPONENTS OF VALUES EQUAL TO 1. FOR LINEAR17
C   DETAILS SEE "INTRODUCTION TO STATISTICAL PATTERN RECOGNITION" LINEAR18
C   BY FUKUNAGA, ACADEMIC PRESS, 1972. LINEAR19
C   THE MATRICES (DIMENSIONED AS ONE DIMENSIONAL IN PG4) HAVE THE LINEAR20
C   FOLLOWING STRUCTURES--            LINEAR21
C   XTRAIN --- XTRAIN IS AN NDIM X NTRAIN MATRIX. EACH COLUMN LINEAR22
C   VECTOR IN XTRAIN REPRESENTS ONE TRAINING PATTERN. LINEAR23
C   THE FIRST ENTRY IN EACH COLUMN IS THE CATEGORY NAME LINEAR24
C   THE SECOND ENTRY IS THE CONSTANT 1. THE ENTRIES 3 LINEAR25
C   TO NDIM ARE THE VALUES OF THE COMPONENTS OF THE LINEAR26
C   PATTERN VECTOR X(J). XTRAIN HAS NTRAIN COLUMNS AND LINEAR27
C   ND+1 = NDIM ROWS.                LINEAR28
C                                                       LINEAR29
C   XTEST --- XTEST HAS THE SAME CONFIGURATION AS XTRAIN. LINEAR30
C                                                       LINEAR31
C   U --- DATA MATRIX U IS USED FOR CALCULATING THE BOUNDARY LINEAR32
C   BETWEEN THE CATEGORIES I AND J. U HAS ND ROWS AND LINEAR33
C   NTR(I)+NTR(J) COLUMNS, AS FOLLOWS--- LINEAR34
C   ***PATTERNS FROM CATEGORY I*****PATTERNS FROM CATEGORY J -*** LINEAR35
C   1      1      .....      -1      -1      .....      -1 LINEAR36
C   .      .      .....      .      .      .....      . LINEAR37
C   .      .      .....      .      .      .....      . LINEAR38
C   X      X      .....      -X      -X      .....      -X LINEAR39
C   .      .      .....      .      .      .....      . LINEAR40
C   .      .      .....      .      .      .....      . LINEAR41
C   ***** U HAS ND ROWS AND NTR(I)+NTR(J) COLUMNS ***** LINEAR42
C                                                       LINEAR43
C   ENTRY POINT.                       LINEAR44
C   CALL LINEAR(XTRAIN,XTEST,W,U,DUMMY) LINEAR45
C                                                       LINEAR46
C   INPUT ARGUMENTS.                   LINEAR47
C   XTRAIN          MATRIX CONTAINING TRAINING PATTERNS LINEAR48
C   YTEST           MATRIX CONTAINING TEST PATTERNS LINEAR49
C   W               WEIGHT VECTORS LINEAR50
C                                                       LINEAR51

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	I1=(I-1)*NDIM+1		LINEAR201
C		SKIP IF WE HAVE THIS CATEGORY NAME	LINEAR202
	IF(NAMF(J).EQ.XTRAIN(I1)) GO TO 2	OR, SET UP A NEW CATEGORY	LINEAR203
C			LINEAR204
	J=J+1	GET THE NEW NAME	LINEAR205
C			LINEAR206
	NAME(J)=XTRAIN(I1)	COUNT THE NUMBER IN THIS CATEGORY	LINEAR207
C			LINEAR208
2	NTR(J)=NTR(J)+1		LINEAR209
3	CONTINUE	WRITE OUT NO. OF TRAINING AND TEST	LINEAR210
C		PATTERNS AND THE NO. OF CATEGORIES	LINEAR211
C		FOUND	LINEAR212
C			LINEAR213
	WRITE(6,1403) NTRAIN,NTEST,J		LINEAR214
1403	FORMAT(10X,'NUMBER OF TRAINING PATTERNS USED=',I4/10X,'NUMBER OF		LINEAR215
	TEST PATTERNS USED=',I4/10X,'NUMBER OF GROUND TRUTH LABELS FOUND='		LINEAR216
	1,I3)		LINEAR217
C		WRITE OUT THE NO. OF CATS SPECIFIED	LINEAR218
C		BY THE USER	LINEAR219
	WRITE(6,1404) NC		LINEAR220
1404	FORMAT(/10X,'NUMBER OF GROUND TRUTH LABELS SPECIFIED =',I3)		LINEAR221
	WRITE(6,1411)		LINEAR222
1411	FORMAT(1H1,5X,'SUMMARY OF WEIGHT VECTORS'///)		LINEAR223
C			LINEAR224
C		FIND THE HYPERPLANES WHICH SEPARATE	LINEAR225
C		CATEGORIES I AND J, WHERE I GOES FROM	LINEAR226
C		1 TO NC AND J GOES FROM I+1 TO NC.	LINEAR227
C			LINEAR228
	IWCOL=0		LINEAR229
	DO 10 I=1,NC		LINEAR230
	IPLUS1=I+1		LINEAR231
	DO 10 J=IPLUS1,NC	CHECK FOR FINAL LOOP	LINEAR232
C			LINEAR233
	IF(J.GT.NC) GO TO 868	INCREMENT COLUMN INDEX	LINEAR234
C			LINEAR235
	IWCOL=IWCOL+1		LINEAR236
	ISUM=0	SET UP MATRIX U FOR CAT I AND J	LINEAR237
C			LINEAR238
	DO 9 K=1,I		LINEAR239
9	ISUM=ISUM+NTR(K)	FIND THE COL IN XTRAIN WHERE THE	LINEAR240
C		PATTERNS THAT BELONG TO I AND J BEGIN	LINEAR241
C			LINEAR242
	IWBEG=ISUM-NTR(I)		LINEAR243
	ISUM=0		LINEAR244
	DO 8 K=1,J		LINEAR245
8	ISUM=ISUM+NTR(K)		LINEAR246
	JWBEG=ISUM-NTR(J)		LINEAR247
C			LINEAR248
C		GET ENTRIES FROM XTRAIN AND PUT INTO	LINEAR249
C		THE MATRIX U	LINEAR250

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## LINEAR DISCRIMINANT FUNCTION

C	NI=NTR(I)		LINEAR51
	NJ=NTR(J)		LINEAR52
	NIJ=NTR(I)+NTR(J)		LINEAR53
C	DO 7 NCOLU=1,NI	GO THRU EACH TRAINING PATTERN IN CAT I	LINEAR54
C		INDEX COL FOR XTRAIN	LINEAR55
C	NCOLXT=NCOLU+IWBFG		LINEAR56
C	DO 6 NROWU=1,ND	GO THRU EACH COMP IN VECTOR	LINEAR57
C		SET UP INDEXS	LINEAR58
	NPOWX=NPOWU+1		LINEAR59
	NRCU=(NCOLU-1)*ND+NPOWU		LINEAR60
	NRCX=(NCOLXT-1)*NDIM+NROWX		LINEAR61
C	U(NRCU)=XTRAIN(NRCX)	TRANSFER PATTERN TO U FROM XTRAIN	LINEAR62
6	CONTINUE		LINEAR63
C		DO THE SAME FOR CATEGORY J	LINEAR64
C		GO THRU EACH PATTERN IN CAT J	LINEAR65
C	DO 19 NN=1,NJ		LINEAR66
C		SET UP COL INDEX	LINEAR67
	NCOLU=NI+NN		LINEAR68
	NCOLXT=JWSEG+NN		LINEAR69
	DO 18 NROWU=1,ND		LINEAR70
C		SET UP ROW INDEX	LINEAR71
	NROWX=NROWU+1		LINEAR72
	NRCU=(NCOLU-1)*ND+NROWU		LINEAR73
	NRCX=(NCOLXT-1)*NDIM+NROWX		LINEAR74
C		TRANSFER THE VECTOR IN XTRAIN TO U	LINEAR75
C		AND CHANGE THE SIGN	LINEAR76
18	U(NRCU)=(-1.0)*XTRAIN(NRCX)		LINEAR77
19	CONTINUE		LINEAR78
C		CALL THE REGRESSION ROUTINE	LINEAR79
C	CALL WEIGHT(U,DUMMY,WT,ND,NIJ)	GET WEIGHT VECTOR FOR CATS I AND J	LINEAR80
C			LINEAR81
	DO 40 IROWW=1,ND		LINEAR82
C		INDEX THE VECTOR INTO W ARRAY	LINEAR83
	IRCW=(IWCOL-1)*ND+IROWW		LINEAR84
40	W(IRCW)=WT(IROWW)		LINEAR85
C		WRITE OUT THE WEIGHT VECTOR	LINEAR86
	WRITE(6,100) I,J		LINEAR87
	IPCW1=(IWCOL-1)*ND+1		LINEAR88
	IPCW2=IWCOL*ND		LINEAR89
100	WRITE(6,101) (W(IRCW),IRCW=IPCW1,IPCW2)		LINEAR90
	FORMAT(2X,44#THE WEIGHT VECTOR FOR SEPARATING CATEGORIES ,I3,3HAND		LINEAR91
101	,I3)		LINEAR92
	FORMAT((1X,10F12.5))		LINEAR93
C		PUNCH OUT THE WEIGHT VECTOR	LINEAR94
	WRITE(43,100) I,J		LINEAR95
	WRITE(43,1761) (W(IPCW),IRCW=IPCW1,IPCW2)		LINEAR96
			LINEAR97
			LINEAR98
			LINEAR99
			LINEAR00





C		RESET MAX COUNT	LINEAR0
	IF (KOUNT.GT.KMAX) KMAX=KOUNT		LINEAR0
65	CONTINUE		LINEAR0
C		GET THE RESULTS AND OUTPUT THEM	LINEAR0
C			LINEAR0
	IIII=(III-1)*NDIM+1		LINEAR0
	IF (INDEX.FQ.1) TCAT=XTRAIN(IIII)		LINEAR0
	IF (INDEX.FQ.2) TCAT=DUMMY1		LINEAR0
C		WRITE RESULTS FOR THIS PATTERN VECTOR	LINEAR1
	IF ((INDEX.EQ.1).AND.(NOPT3.NE.0).AND.(KMAX.EQ.0)) WRITE (6,1501) IIII		LINEAR1
	1,TCAT		LINEAR1
	IF ((INDEX.EQ.2).AND.(NOPT4.NE.0).AND.(KMAX.EQ.0)) WRITE (6,1502) IIII		LINEAR1
	1,TCAT		LINEAR1
	IF ((INDEX.EQ.1).AND.(NOPT3.NE.0).AND.(KMAX.GT.0)) WRITE (6,1502) IIII		LINEAR1
	1,TCAT,NAME(ICLASS),KMAX		LINEAR1
	IF ((INDEX.EQ.2).AND.(NOPT4.NE.0).AND.(KMAX.GT.0)) WRITE (6,1502) IIII		LINEAR1
	1,TCAT,NAME(ICLASS),KMAX		LINEAR1
1501	FORMAT(6X,I3,11X,A5,10X,14HNOT CLASSIFIED)		LINEAR1
1502	FORMAT(6X,I3,11X,A5,10X,A5,12X,I3)		LINEAR2
	DO 71 ITCAT=1,NC		LINEAR2
	IF (TCAT.EQ.NAME(ITCAT)) JTCAT=ITCAT		LINEAR2
71	CONTINUE		LINEAR2
	ITCAT=JTCAT		LINEAR2
C		UPDATE THE CONTINGENCY TABLE	LINEAR2
72	IF (KMAX.GT.0) IERROR(ITCAT,ICLASS)=IERROR(ITCAT,ICLASS)+1		LINEAR2
66	CONTINUE		LINEAR2
C		WRITE OUT THE HEADING	LINEAR2
	IF (INDEX.EQ.1) WRITE (6,721)		LINEAR2
	IF (INDEX.EQ.2) WRITE (6,722)		LINEAR3
721	FORMAT(1H1,15X,'CONTINGENCY TABLE FOR TRAINING PATTERNS')		LINEAR3
722	FORMAT(1H1,15X,'CONTINGENCY TABLE FOR TEST PATTERNS')		LINEAR3
	WRITE (6,723)		LINEAR3
723	FORMAT(4X,'TRUE'/2X,'CATEGORY')		LINEAR3
C		WRITE OUT THE CONTINGENCY TABLE	LINEAR3
	DO 74 ITCAT=1,NC		LINEAR3
74	WRITE (6,724) NAME(ITCAT),(IERROR(ITCAT,ICLS),ICLS=1,NC)		LINEAR3
724	FORMAT(//3X,A6,15(4X,I3))		LINEAR3
C		LIST THE CATEGORIES	LINEAR3
	WRITE (6,725) (NAME(ICLA),ICLA=1,NC)		LINEAR4
725	FORMAT(//11X,15(1X,A6))		LINEAR4
	WRITE (6,726)		LINEAR4
726	FORMAT(//15X,'ASSIGNED CATEGORY')		LINEAR4
67	CONTINUE		LINEAR4
	RETURN		LINEAR4
	END		LINEAR4



CWEIGHT

GET THE WEIGHT VECTOR

WEIGHT0:

WRITTEN BY SAM SHANMUGAM  
DOCUMENTED BY RJ BOSLEY

SEPT 1972  
DEC 1973

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DESCRIPTION OF PROGRAM.

GIVEN A MATRIX U CONTAINING PATTERNS OF CATEGORY I AND J, THIS ROUTINE FINDS A HYPERPLANE TO SEPARATE THESE PATTERNS. THE WEIGHT VECTOR DEFINING THE HYPERPLANE IS GIVEN BY

$$WT = ((UU^T)^{-1}) * (U^T)$$

WHERE WT IS THE WEIGHT VECTOR  
T IS A COLUMN VECTOR WHOSE COMPONENTS ARE +1'S  
ND IS THE LENGTH OF THE PATTERN VECTORS, AND  
NIJ IS THE NUMBER OF PATTERNS FROM CAT I + NUMBER OF PATTERNS FROM CATEGORY J

THE DATA MATRIX U, USED FOR CALCULATING THE BOUNDARY BETWEEN THE CATEGORIES I AND J, IS FORMED AS FOLLOWS ---

\*\*\*PATTERNS FROM CATEGORY I\*\*\*\*\*PATTERNS FROM CATEGORY J -\*\*\*

1	1	.....	-1	-1	....	-1
.	.	.....	.	.	....	.
.	.	.....	.	.	....	.
X	X	.....	-X	-X	....	-X
.	.	.....	.	.	....	.
.	.	.....	.	.	....	.

\*\*\*\*\* U HAS ND ROWS AND NIJ COLUMNS. \*\*\*\*\*

ENTRY POINT.

CALL WEIGHT(U,DUMMY,WT,ND,NIJ)

INPUT ARGUMENTS.

U MATRIX CONTAINING PATTERNS OF CATEGORIES I AND J  
DUMMY DUMMY ARRAY USED FOR MINV  
ND LENGTH OF PATTERN VECTORS IN U  
NIJ NUMBER OF PATTERNS FROM CAT I AND J

OUTPUT ARGUMENT.

WT WEIGHT VECTOR DEFINING THE HYPERPLANE SEPARATING PATTERNS OF CAT I AND J

SUBPROGRAMS REQUIRED.

MINV MATRIX INVERSION PGM FROM IBM SSP

INTERNAL PARAMETERS.

L,M SCRATCH ARRAYS USED BY MINV  
D DETERMINANT OF DUMMY, FROM MINV  
IROWW ROW, OR COMPONENT, INDEX FOR WT  
IDJD INDEX FOR A VECTOR IN DUMMY  
IROWWK INDEX FOR A VECTOR IN DUMMY AFTER IT

07-09-74 20.434

GET THE WEIGHT VECTOR

```
C                                     HAS BEEN INVERTED BY MINV
C
C
C SUBROUTINE WEIGHT(U,DUMMY,WT,ND,NIJ)
C DIMENSIC : U(1),DUMMY(1),WT(100)
C DIMENSION L(100),M(100)
C                                     SET UP MATRIX DUMMY=U*U*
C                                     GO THRU EACH COMPONENT
C
C DO 15 ID=1,ND
C DO 15 JD=1,ND
C SUM=0.0
C                                     GO THRU EACH VECTOR
C
C DO 14 K=1,NIJ
C IDK=(K-1)*ND+ID
C JDJ=(K-1)*ND+JD
C
C FIND U*U*
C 14 SUM=SUM+U(IDK)*U(JDK)
C IDJD=(JD-1)*ND+ID
C
C STORE IT IN DUMMY FOR INVERSION
C
C DUMMY(IDJD)=SUM
C 15 CONTINUE
C
C FIND THE INVERSE OF U*U*
C CALL MINV(DUMMY,ND,0,L,M)
C DO 30 K=1,ND
C SUM=0.0
C DO 25 KK=1,NIJ
C
C COMPUTE U*T
C
C 25 KKK=(KK-1)*ND+K
C SUM=SUM+U(KKK)
C
C STORE IT IN U(K,1),K=1,ND
C
C 30 U(K)=SUM
C
C COMPUTE THE WEIGHT VECTOR
C W(IROWW,NCOL),IROWW=1,ND
C
C DO 40 IROWW=1,ND
C SUM=0.0
C DO 35 K=1,ND
C IROWWK=(K-1)*ND+IROWW
C 35 SUM=SUM+DUMMY(IROWWK)*U(K)
C
C STORE IT IN WT ARRAY AND RETURN
C
C 40 WT(IROWW)=SUM
C RETURN
C END
```

WEIGHT51  
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APPENDIX V

## Glossary and Index to Remotely Sensed Image Pattern Recognition Concepts\*

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(Received 29 March 1973 and in revised form 21 June 1973)

**Abstract**—The purpose of the glossary is to state in the simplest possible way the general meaning or word usage for many of the terms in image pattern recognition. There is no intent to provide definitive statements for terms such as "resolution" but rather only statements about the general nature of what resolution is. There is no intent to provide mathematical formulas involving integrals or derivatives in any of the statements. Those who need the mathematics can get it from technical papers or texts.

The glossary is designed to be read by those generally unfamiliar with the area and provide for them an overall perspective. The organization approaches that of programmed learning material and can be smoothly (I hope) read from beginning to end. Those needing to look up a specific term can do so via the index.

There is some overlap of terms in this glossary with those glossaries or definitions in radiometry and aerial photography. There is no intent that the way the terms are described here replace the way they are described in those glossaries and definitions. The overlap is provided here so that the reader can get a perspective of a cluster of terms frequently used in our field. The perspective is intended to start from what the image concept is through the recording of an image by some sensor, the possible conversion of image format and the simple analog or more complex digital processing which must be done on the imagery. In short, the perspective is one of image pattern recognition.

1. An *Image* is a spatial representation of an object, scene, or another image. It can be real or virtual as in optics. In pattern recognition, image usually means a recorded image such as a photograph, map, or picture. It may be abstractly thought of as a continuous function  $I$  of two variables defined on some bounded region of a plane. When the image is a photograph, the range of the function  $I$  is the set of grey shades usually considered to be normalized to the interval  $[0, 1]$ . The grey shade located at spatial coordinate  $(x, y)$  is denoted by  $I(x, y)$  and is usually proportional to the radiant energy in the electromagnetic band to which the photographic sensor is sensitive. When the image is a map, the range of the function  $I$  is a set of symbols or colors, and the symbol or color located at spatial coordinate  $(x, y)$  is denoted by  $I(x, y)$ . A recorded image may be in photographic, video signal, or digital format.

2. The *grey shade* or *grey tone* is a number or value assigned to a position  $(x, y)$  on an image. The number is proportional to the integrated output, reflectance, or transmittance of a small area, usually called a resolution cell or pixel, centered on the position  $(x, y)$ . The grey shade can be measured as or expressed in any one of the following ways:

- (1) transmittance
- (2) reflectance
- (3) a coordinate of the ICI color coordinate system

\* The glossary was prepared as the report from the definitions and standards subcommittee, Automatic Image Pattern Recognition Committee of Electronic Industries Association.

† I would like to acknowledge the helpful suggestions, comments and corrections which many of my colleagues made. Particular thanks are due to R. Asendorf, B. Bulluck, L. Kirvida, P. Pryor, A. Rosenfeld, B. Scheps, R. Swonger, S. Viglione and R. Zaputowycz.

- (4) a coordinate of the tristimulus value color coordinate system
- (5) brightness
- (6) radiance
- (7) luminance
- (8) density
- (9) voltage
- (10) current.

3. A *photograph* is a "hard copy" pictorial record of an image formed by a sensor. The photograph is usually recorded on some type of photosensitive emulsion. It can be either reflective, as is a paper print, or transmissive, as is a transparency. It is usually two-dimensional and its reflectance or transmittance, (either monochromatic or polychromatic) varies as a function of position. If it is a multi-colored image (polychromatic), it can be either natural color where the colors are similar to the original, or false color where the colors of the photograph are radically different from the original. The sensor used to form the image may be any type such as an optical camera with or without spectral filtration, infrared optical-mechanical scanners, TV systems, radars, or sonic sensors, etc. The type of sensor recording the image and spectral region the sensor is sensitive to, should always be indicated when referring to a photograph.

4. A *map* is a representation, of physical and/or cultural features (natural, artificial or both) of a region (such as the sky) or a surface such as that of the earth or a planet. It indicates by a combination of symbols and colors those regions having designated category identifications. Very often ground truth and/or decision rule category assignments are displayed by maps. A photograph with limited symbolism and annotation is often called a *photo-map*.

5. The *radiant intensity* of a point object is a measure of the radiant power per steradian radiated or reflected by an object. In general, radiant intensity is a function of the nature of the object, the viewing angle, spectral wavelength and band-width.

6. The *reflectance* or *reflection coefficient* is the ratio of the energy per unit time per unit area (radiant power density) reflected by the object to the energy per unit time per unit area incident on the object. In general, reflectance is a function of the incident angle of the energy, viewing angle of the sensor, spectral wavelength and bandwidth, and the nature of the object.

7. The *transmittance* or *transmittance coefficient* is the ratio of the energy per unit time per unit area (radiant power density) transmitted through the object to the energy per unit time per unit area incident on the object. In general, transmittance is a function of the incident angle of the energy, viewing angle of the sensor, spectral wavelength and band-width, and the nature of the object.

8. The *density* of an  $(x, y)$  position on a photograph is a measure of the light absorbing capability of the silver or dye deposited on that position. It is defined by the logarithm of the position's reciprocal transmittance. The density measured should be specified as to whether it is specular or diffuse.

9. *Densitometry* is the field devoted to the measurement of optical image densities on film or print grey shades usually caused by the absorption or reflection of light by developed photographic emulsion.

10. A *densitometer* is a device used to measure the average image density of a small area of specified size on a photographic transparency or print. The measurement may be a

meter reading or an electronic signal. When the small area is smaller than a few hundred microns square, the instrument is called a *micro-densitometer*.

11. The *contrast* for a point object against its background can be measured by: (1) its *contrast ratio*, which is the ratio between the higher of object transmittance or background transmittance to the lower of object transmittance or background transmittance; (2) its *contrast difference*, which is the difference between the higher density of object or background to the lower density of object or background; (3) its *contrast modulation*, which is the difference between the darker of object or background grey shade and the lighter of object or background grey shade divided by the sum of object grey shade and background grey shade.

12. *Resolution* is a generic term which describes how well a system, process, component or material, or image can reproduce an isolated object or separate closely spaced objects or lines. The *limiting resolution*, *resolution limit* or *spatial resolution* is described in terms of the smallest dimension of the target or object that can just be discriminated or observed. Resolution may be a function of object contrast, spatial position as well as element shape (single point, number of points in a cluster, continuum, or line etc.).

13. The *resolving power* of an imaging system, process, component or material is a measure of its ability to image closely spaced objects. The most common practice in measuring resolving power is to image a resolving power target composed of lines and spaces of equal width. Resolving power is usually measured at the image plane in line pairs per millimeter, i.e. the greatest number of lines and spaces per millimeter that can just be recognized. This threshold is usually determined by using a series of targets of decreasing size and basing the measurement on the smallest one in which all lines can be counted. In measuring resolving power the nature of the target (number of lines and their aspect ratio), its contrast and the criteria for determining the limiting resolving power *must be* specified.

14. *Acutance* is a measure of the sharpness of edges in a photograph or image. It is defined for any edge by the average squared rate of change of the density across the edge divided by the total density difference from one side of the edge to the other side of the edge.

15. The *spread function* of an image system, process, component, or material describes the resulting spatial distribution of grey shade when the input to the system is some well defined object much smaller than the width of the spread function. If the input to the system is a line, the spread function is called the *line spread function*. If the input to the system is a point, the spread function is called the *point spread function*.

16. The *Modulation Transfer Function* of an imaging system or component measures the spatial frequency modulation response of the system or component. As an imaging system or component processes or records an image, the contrast modulation of the processed or recorded image is different from the input image. In fact, there is always a spatial frequency beyond which the contrast modulation of the processed or recorded (output) image is smaller (worse) than the contrast modulation of the input image. The modulation transfer function can be thought of as a curve indicating, for each spatial frequency, the ratio of the contrast modulation of the output image to the contrast modulation of the input image. It is formally defined as the magnitude of the Fourier transform of the line spread function of the imaging system or component.

17. A *resolution cell* is the smallest most elementary areal constituent of grey shades considered by an investigator in an image. A resolution cell is referenced by its spatial coordinates. The resolution cell or formations of resolution cells can sometimes constitute the basic unit for pattern recognition of image format data.

18. A *digital image*, or *digitized image*, or *digital picture function* of an image is an image in digital format and is obtained by partitioning the area of the image into a finite two-dimensional array of small uniformly shaped mutually exclusive regions, called resolution cells, and assigning a "representative" grey shade to each such spatial region. A digital image may be abstractly thought of as a function whose domain is the finite two-dimensional set of resolution cells and whose range is the set of grey shades.

19. A *picture element* or *pixel* or *pel* is a pair whose first member is a resolution cell and whose second member is the grey shade assigned by the digital image to that resolution cell. Sometimes picture element, pixel, or pel refer only to the grey shade or grey shade  $n$ -tuple in a resolution cell.

20. A *multi-image* is a set of images, each taken of the same subject at different times, or from different positions, or with different sensors, or at different electromagnetic frequencies, or with different polarizations. Although there is a high degree of information redundancy between images in a multi-image set, each image usually has information not available in any one of or combinations of the other images in the set.

21. A *multi-digital image* is a multi-image in digital form. It can be, for example, a set of digital images obtained from the images in a multi-image. A multi-digital image is often called a multi-image for short when it is understood from context that digital images are involved.

22. A *flying spot scanner* is a device used to rapidly convert image data from photographic format to electronic video signal format. Normally, the scanner directs an electron beam across the face of a cathode ray tube (CRT) in a TV-like raster. The photographic transparency is placed in front of the CRT (either directly or through some optics) and the light coming from the CRT is passed through it. The modulated light beam is detected by a photomultiplier or other photo detector and amplified to a usable video signal level.

23. A *scanning densitometer* is a device used to convert image data from photographic format to electronic video signal format. Usually, the photographic transparency is placed on a glass cylinder which rotates and slowly translates. A fine beam of light is focused on the transparency, passed through it, and is detected by a photo-multiplier where it is amplified to a usable video signal. The scanning densitometer is a much slower conversion device than the flying spot scanner. However, this disadvantage is compensated by its fine resolution capability of a few microns.

24. The *vidicon* is an imaging vacuum tube having a photosensitive surface and is a means of converting image data from instantaneous radiance format to electronic video signal format. The scene being viewed is imaged on the photosensitive surface which can be scanned by an electron beam generating a signal whose amplitude corresponds to the radiant intensity focused on the surface at each point. This signal is called a *video signal* and may be amplified to any desired level.

25. A *video image* is an image in electronic signal format capable of being displayed on a cathode ray tube screen. The video signal is generated from devices like a vidicon or flying spot scanner which converts an image from photographic form to video signal form by scanning it line by line. The video signal itself is a sequence of signals, the  $i$ th signal representing the  $i$ th line of the scanned image.

26. *Registering* is the translation-rotation alignment process by which two images of like geometries and of the same set of objects are positioned coincident with respect to one another so that corresponding elements of the same ground area appear in the same place on the registered images. In this manner, the corresponding grey shades of the two images

at any  $(x, y)$  coordinate or resolution cell will represent the sensor output for the same object over the full image frame being registered.

27. *Congruencing* is the process by which two images of a multi-image set are transformed so that the size and shape of any object on one image is the same as the size and shape of that object on the other image. In other words, when two images are congruenced, their geometries are the same and they coincide exactly.

28. *Rectifying* is a process by which the geometry of an image area is made planimetric. For example, if the image is taken of an equally spaced rectangular grid pattern, then the rectified image will be an image of an equally spaced rectangular grid pattern. Rectification does not remove relief distortion.

29. *Change detection* is the process by which two images may be compared, resolution cell by resolution cell, and an output generated whenever corresponding resolution cells have different enough grey shades or grey shade  $n$ -tuples.

30. An *optical color combiner* is an instrument which produces "false" or "true" color images by linearly combining a few black and white transparencies of the same scene. The transparencies are usually obtained from multi-spectral, multi-band, or time-sequential photography. The transparencies are placed in projectors which are all focused and registered on the same screen and which have various color filters placed in front of their lenses. The viewing brightness of the projector's lamp in each projector can be changed independently thereby changing chromaticity balance. An optical color combiner is sometimes called an *additive color display*.

31. An *electronic color combiner* is an instrument which produces a "false" color image by linearly combining video signals of images of the same scenes. The images are usually obtained from multi-spectral, multi-band, or time-sequential photography. If the original image format is photographic, then the image format is changed from photographic to video signal format by synchronized vidicons or flying spot scanners. The resulting video signals are linearly combined through a matrix multiplier circuit, and the three linearly combined signals then drive the color gun of a color TV tube. An electronic color combiner usually has greater versatility for congruencing or registering than an optical color combiner.

32. *Level slicing* or *density slicing* or *thresholding* is an operation performed by an instrument (usually electronic) called a level slicer to change one or more a grey scale images to one binary image.

33. The *level slicer*, *density slicer* or *thresolder* is an instrument (usually electronic) which takes a single or multi-image as an input and produces a binary image for an output. A binary "one" is produced on the output image whenever the grey shades on each of the input images lie within the independently set minimum and maximum thresholds. A set of  $N$  input images would, therefore, require a setting  $N$  minimum and  $N$  maximum levels.

34. A *figure  $F$* , or a *subimage  $F$*  in a continuous or digital image  $I$  is any function  $F$  whose domain is some subset  $A$  of the set of spatial coordinates or resolution cells, whose range is the set  $G$  of grey shades, and which is defined by  $F(x, y) = I(x, y)$  for any  $(x, y)$  belonging to  $A$ .

35. A figure  $F$  is *connected* if there is a path between any two spatial coordinates or resolution cells contained in the domain of  $F$ . More precisely,  $F$  is connected if for each pair of spatial coordinates  $(x, y)$  and  $(u, v)$  belonging to the domain of  $F$ , there exists some sequence  $\langle (a_1, b_1), (a_2, b_2), \dots, (a_m, b_m) \rangle$  of spatial coordinates belonging to the domain



of  $F$  such that  $(x, y) = (a_1, b_1)$ ,  $(u, v) = (a_m, b_m)$ , and  $(a_i, b_i)$  and  $(a_{i+1}, b_{i+1})$  are sufficiently close neighboring coordinates,  $i = 1, 2, \dots, m-1$ .

36. A figure  $F$  is *convex* if the domain of  $F$  contains the line segment which joins any pair of spatial coordinates in the domain of  $F$ .

37. A *discrete tonal feature* on a continuous or digital image is a connected set of spatial coordinates or resolution cells all of which have the same or almost the same grey shade.

38. *Texture* is concerned with the spatial distribution of the grey shades and discrete tonal features. When a small area of the image has little variation of discrete tonal features, the dominant property of that area is grey shade. When a small area has wide variation of discrete tonal features, the dominant property of that area is texture. There are three things crucial in this distinction: (1) the size of the small areas, (2) the relative sizes of the discrete tonal features, and (3) the number of distinguishable discrete tonal features.

39. *Quantizing* is the process by which each grey shade in an image of photographic, video, or digital format is assigned a new value from a given finite set of grey shade values. There are three often used methods of quantizing:

(1) in *equal interval quantizing* or *linear quantizing*, the range of grey shades from maximum grey shade to minimum grey shade is divided into contiguous intervals each of equal length, and each grey shade is assigned to the quantized class which corresponds to the interval within which it lies;

(2) in *equal probability quantizing*, the range of grey shades is divided into contiguous intervals such that after the grey shades are assigned to their quantized class there is an equal frequency of occurrence for each quantized grey shade in the quantized digital image or photograph; equal probability quantizing is sometimes called *central stretching*;

(3) in *minimum variance quantizing*, the range of grey shades is divided into contiguous intervals such that the weighted sum of the variance of the quantized intervals is minimized. The weights are usually chosen to be the grey shade interval probabilities which are computed as the proportional area on the photograph or digital image which have grey shades in the given interval.

40. A *quantizer* is an instrument which does quantizing. The quantizer has three functional parts. The first part allows the determining and/or setting of the quantizing intervals, the second part is a level slicer which indicates when a signal is in any quantizing interval, and the third part takes the binary output from the level slicers and either codes it to some binary code or converts it to some analog signal representing quantizing interval centers or means.

41. The simplest and most practical *unit* to observe and measure in the pattern recognition of image data is often the basic picture element (the grey shade or the grey shade  $n$ -tuple in its particular resolution cell). This is what makes pattern recognition so hard sometimes for the objects requiring analysis or identification are not simple picture elements but are often complex spatial formations of picture elements such as houses, roads, forest, etc.

42. A *measurement  $n$ -tuple* or *measurement pattern* or *pattern* or *measurement vector* is the ordered  $n$ -tuple of measurements obtained of a unit under observation. Each component of the  $n$ -tuple is a measurement of a particular quality, property, feature, or characteristic of the unit. In image pattern recognition, the units are usually picture elements or simple formations of picture elements and the measurement  $n$ -tuples are the corresponding grey shades, grey shade  $n$ -tuples, or formations of grey shade  $n$ -tuples.

43. The *range set*  $R_i$  for the  $i$ th sensor which produces the  $i$ th image in the multi-image set, is the set of all measurements which can be produced by the  $i$ th sensor. Simply, it is the set of all grey shades which could possibly exist on the  $i$ th image.

44. The *Cartesian product* of two sets  $A$  and  $B$ , denoted by  $A \times B$ , is the set of all ordered pairs where the first component of the pair is some element from the first set and the second component of the pair is some element from the second set. The Cartesian product of  $N$  sets can be inductively defined in the usual fashion.

45. *Measurement space* is a set large enough to include in it the set of all possible measurement  $n$ -tuples which could be obtained by observing physical attributes of some set of units. When the units are single resolution cells or picture elements, measurement space  $M$  is the Cartesian product of the range sets of the sensors;  $M = R_1 \times R_2 \times \dots \times R_n$ .

46. Each unit is assumed to be of one and only one given type. The set of types is called the set of *pattern classes* or *categories*  $C$ , each type being a particular category. The categories are chosen specifically by the investigator as being the ones of interest to him.

47. A *feature* or *feature pattern* or *feature  $n$ -tuple* or *pattern feature* is a  $n$ -tuple or vector with (a small number of) components which are functions of the initial measurement pattern variables or some subsequence of the measurement  $n$ -tuples. Feature  $n$ -tuples or vectors are designed to contain a high amount of information relative to the discrimination between units of the types of categories in the given category set. Sometimes the features are predetermined and other times they are determined at the time the pattern discrimination problem is being solved. In image pattern recognition, features often contain information relative to grey shade, texture, shape or context.

48. *Feature space* is the set of all possible feature  $n$ -tuples.

49. *Feature selection* is the process by which the features to be used in the pattern recognition problem are determined. Sometimes feature selection is called *property selection*.

50. *Feature extraction* is the process in which an initial measurement pattern or some subsequence of measurement patterns is transformed to a new pattern feature. Sometimes feature extraction is called *property extraction*.

51. The word *pattern* can be used in three distinct senses:

(1) as measurement pattern;

(2) as feature pattern; and

(3) as the dependency pattern or patterns of relationships among the components of any measurement  $n$ -tuple or feature  $n$ -tuple derived from units of a particular category and which are unique to those  $n$ -tuples, that is, they are dependencies which do not occur in any other category.

52. A *signature* is the observable or characteristic measurement or feature pattern derived from units of a particular category. A category is said to have a signature only if the characteristic pattern is highly representative of the  $n$ -tuples obtained from units of that category. Sometimes a signature is called a *prototype pattern*.

53. A *data sequence*  $S_d = \langle d_1, d_2, \dots, d_j \rangle$  is a sequence of patterns derived from the measurement patterns or features of some sequence of observed units.  $d_1$  is the pattern associated with the first unit;  $d_2$  is the pattern associated with the second unit; and  $d_j$  is the pattern associated with the  $j$ th unit.

54. A *decision rule*  $f$  usually assigns one and only one category to each observed unit on the basis of the sequence of measurement patterns in the data sequence  $S_d$  or in the corresponding sequence of feature patterns.

55. A *simple decision rule* is a decision rule which assigns a unit to a category solely on the basis of the measurements or features associated with the unit. Hence, the units are treated independently and the decision rule  $f$  may be thought of as a function which assigns one and only one category to each pattern in measurement space or to each feature in feature space.

56. A *compound decision rule* is a decision rule which assigns a unit to a category on the basis of some non-trivial subsequence of measurement patterns in the data sequence or in the corresponding sequence of feature patterns. A compound decision rule is not a simple decision rule.

57. Provision can be made for the decision rule to *reserve judgement* or to *defer assignment* if the pattern is too close to the category boundary in measurement or feature space. With this provision, a deferred assignment is an assignment to the category of "reserved judgement."

58. A *category identification sequence* or *ground truth*  $S_c = \langle c_1, c_2, \dots, c_J \rangle$  is a sequence of category identifications obtained from some sequence of observed units.  $c_1$  is the category identification of the first unit;  $c_2$  is the category identification of the second unit; and  $c_J$  is the category identification of the  $J$ th unit.

59. A *training sequence* is a set of two sequences: (1) the data sequence and (2) a corresponding category identification sequence (sometimes called ground truth). The training sequence is used to estimate the category conditional probability distributions from which the decision rule is constructed.

60. The *conditional probability* of a measurement or feature  $n$ -tuple  $d$  given category  $c$  is denoted by  $P_c(d)$ , or by  $P(d/c)$ , and is defined as the relative frequency or proportion of times the  $n$ -tuple  $d$  is derived from a unit whose true category identification is  $c$ .

61. A *distribution-free* or *non-parametric decision rule* is one which makes no assumptions about the functional form of the conditional probability distribution of the patterns given the categories.

62. A simple *maximum likelihood decision rule* is one which treats the units independently and assigns a unit  $u$  having pattern measurement or features  $d$  to that category  $c$  whose units are most probable to have given rise to pattern or feature vector  $d$ , that is, such that the conditional probability of  $d$  given  $c$ ,  $P_c(d)$ , is highest.

63. A simple *Bayes decision rule* is one which treats the units independently and assigns a unit  $u$  having pattern measurements or features  $d$  to the category  $c$  whose conditional probability,  $P_d(c)$ , given measurement  $d$ , is highest.

64. Let  $\langle u_1, u_2, \dots, u_J \rangle$  be a sequence of units with corresponding data sequence  $\langle d_1, d_2, \dots, d_J \rangle$  and known category identification sequence  $\langle c_1, c_2, \dots, c_J \rangle$ . A simple *nearest neighbor decision rule* is one which treats the units independently and assigns a unit  $u$  of unknown identification and with pattern measurements or features  $d$  to category  $c_j$  where  $d_j$  is that pattern closest to  $d$  by some given metric or distance function.

65. A *discriminant function*  $f_i(d)$  is a scalar function, whose domain is usually measurement space and whose range is usually the real numbers. When  $f_i(d) \geq f_k(d)$ ,  $k=1, 2, \dots, K$ , then the decision rule assigns the  $i$ th category to the unit giving rise to pattern  $d$ .

66. A *linear discriminant function*  $f$  is a discriminant function of the form

$$f(d) = b + \sum_{j=1}^n a_j \delta_j \text{ where } d = (\delta_1, \delta_2, \dots, \delta_n).$$

67. A *decision boundary* between the  $i$ th and  $k$ th categories is a subset  $H$  of patterns in measurement space  $M$  defined by

$$H = \{d \in M | f_i(d) = f_k(d)\},$$

where  $f_i$  and  $f_k$  are the discriminant functions for the  $i$ th and  $k$ th categories.

68. A *hyperplane decision boundary* is the special name given to decision boundaries arising from the use of linear discriminant functions.

69. A *linear decision rule* is a simple decision rule which usually treats the units independently and makes the category assignments using linear discriminant functions. The decision boundaries obtained from linear decision rules are hyperplanes.

70. The *pattern discrimination* problem is concerned with how to construct the decision rule which assigns a unit to a particular category on the basis of the measurement pattern(s) in the data sequence or on the basis of the feature pattern(s) in the data sequence.

71. *Pattern identification* is the process in which a decision rule is applied. If  $S_u = \langle u_1, u_2, \dots, u_J \rangle$  is the sequence of units to be observed and identified, and if  $S_d = \langle d_1, d_2, \dots, d_J \rangle$  is the corresponding data sequence of patterns, then the pattern identification process produces a category identification sequence  $S_c = \langle c_1, c_2, \dots, c_J \rangle$  where  $c_i$  is the category in  $C$  to which the decision rule assigns unit  $u_i$  on the basis of the  $J$  patterns in  $S_d$ . In general, each category in  $S_c$  can be assigned by the decision rule as a function of all the patterns in  $S_d$ . Sometimes pattern identification is called "*pattern classification*" or "*classification*".

72. A *cluster* is a homogenous group of units which are very "like" one another. "Likeness" between units is usually determined by the association, similarity, or distance between the measurement patterns associated with the units.

73. A *cluster assignment function* is a function which assigns each observed unit to a cluster on the basis of the measurement pattern(s) in the data sequence or on the basis of their corresponding features. Sometimes the units are treated independently; in this case the clustering assignment function can be considered as a transformation from measurement space to the set of clusters.

74. The *pattern classification* problem is concerned with constructing the cluster assignment function which groups similar units. Pattern classification is synonymous with *numerical taxonomy* or *clustering*.

75. The *cluster identification* process is the process in which the cluster assignment function is applied to the sequence of observed units thereby yielding a cluster identification sequence.

76. A *misidentification*, or *misdetection*, or *type I error* occurs for category  $c_i$  if a unit whose true category identification is  $c_i$  is assigned by the decision rule to category  $c_k$ ,  $k \neq i$ . A misidentification error is often called an *error of omission*.

77. A *false identification*, or *false alarm*, or *type II error* occurs for category  $c_i$  if a unit whose true category identification is  $c_k$ ,  $k \neq i$ , is assigned by the decision rule to category  $c_i$ . A false identification error is often called an *error of commission*.

78. A *prediction sequence*, or *test sequence*, or a *generalization sequence* is a set of two sequences: (1) a data sequence (whose corresponding true category identification sequence may be considered to be unknown to the decision rule) and (2) a corresponding category identification sequence determined by the decision rule assignment. By comparing the category identification sequence determined by the decision rule assignment with the

category identification sequence determined by the ground truth, the misidentification rate and the false identification rate for each category may be estimated.

79. A *confusion matrix* or *contingency table* is an array of probabilities whose rows and columns are both similarly designated by category label and which indicates the probability of correct identification for each category as well as the probability of type I and type II errors. The (*i*th, *k*th) element  $P_{ik}$  is the probability that a unit has true category identification  $c_i$ ; and is assigned by the decision rule to category  $c_k$ .

80. A unit is said to be *detected* if the decision rule is able to assign it as belonging only to some given subset *A* of categories from the set *C* of categories. To detect a unit does not imply that the decision rule is able to identify the unit as specifically belonging to one particular category.

81. A unit is said to be *recognized, identified, classified, categorized* or *sorted* if the decision rule is able to assign it to some category from the set of given categories. In military applications, there is a definite distinction between recognize and identify. Here, for a unit to be recognized, the decision rule must be able to assign it to a type of category, the type having included within it many subcategories. For a unit to be identified, the decision rule must be able to assign it not only to a type of category but also to the subcategory of the category type. For example, a small area ground patch may be recognized as containing trees, which may be specifically identified as apple trees.

82. A unit is said to be *located* if specific coordinates can be given for the unit's physical location.

83. A unit is said to be *acquired* if it can be located and recognized.

84. A *target* is one type of category used in the pattern recognition of image data. It usually occupies some relatively small area on the image and has a unique or characteristic set of attributes. It has a high *a priori* interest to the investigator.

85. *Target discrimination* is the process by which decision rules for targets (small area extensive categories) are constructed.

86. *Target identification* or *target recognition* is the process by which targets contained within image data are identified by means of a decision rule.

87. An *image transformation* is a function or operator which takes an image for its input and produces an image for its output. The domain of the transform operator is often called the spatial domain. The range of the transform operator is often called the transformed domain. Some transformations have spatial and transform domains of entirely different character. For these transforms, the image in the spatial domain may appear entirely different from and have a different interpretation from the image in the transformed domain. Specific examples of these kinds of transformations are the Fourier, Hadamard, and Karhunen-Loève transformations. Other transformations have spatial and transform domain of similar character. For these transformations, the image in the transformed domain may appear similar to the image of the spatial domain. These types of transformations are often called spatial filters.

88. A *spatial filter* is an image transformation, usually a one-one operator used to lessen noise or enhance certain characteristics of the image. For any particular (*x*, *y*) coordinate on the transformed image, the spatial filter assigns a grey shade on the basis of the grey shades of a particular spatial pattern near the coordinates (*x*, *y*).

89. A *linear spatial filter* is a spatial filter for which the grey shade assignment at coordinates (*x*, *y*) in the transformed image is made by some weighted average (linear combination) of grey shades located in a particular spatial pattern around coordinates (*x*, *y*) of the domain image. The linear spatial filter is often used to change the spatial

frequency characteristics of the image. For example, a linear spatial filter which emphasizes high spatial frequencies will tend to sharpen the edges in an image. A linear spatial filter which emphasizes the low spatial frequencies will tend to blur the image and reduce salt and pepper noise.

90. *Template matching* is an operation which can be used to find out how well two photographs or images match one another. The degree of matching is often determined by cross-correlating the two images or by evaluating the sum of the squared corresponding grey shade differences. Template matching can also be used to best match a measurement pattern with a prototype pattern.

91. *Matched filtering* is a template matching operation done by using the magnitude of the cross-correlation function to measure the degree of matching.

92. In pattern recognition problems such as target discrimination, for which the category of interest is some specific formation of resolution cells with characteristic shape or tone-texture composition, the problem of pattern segmentation may occur. *Pattern segmentation* is the problem of determining which regions or areas in the image constitute the patterns of interest, i.e. which resolution cells should be included and which excluded from the pattern measurements.

93. *Screening* is the operation of separating the uninteresting photographs or images from those photographs containing areas of potential interest.

94. *Preprocessing* is an operation applied before pattern identification is performed. Preprocessing produces, for the categories of interest, pattern features which tend to be invariant under changes such as translation, rotation, scale, illumination levels, and noise. In essence, preprocessing converts the measurements patterns to a form which allows a simplification in the processing rule. Preprocessing can bring into registration, bring into congruence, remove noise, enhance images, segment target patterns, detect, center, and normalize targets of interest.

95. *Image compression* is an operation which preserves all or most of the information in the image and which reduces the amount of memory needed to store an image or the time needed to transmit an image.

96. *Image restoration* is a process by which a degraded image is restored to its original condition. Image restoration is possible only to the extent that the degradation transform is mathematically invertible.

97. *Image enhancement* is any one of a group of operations which improve the detectability of the targets or categories. These operations include, but are not limited to, contrast improvement, edge enhancement, spatial filtering, noise suppression, image smoothing, and image sharpening.

98. *Image processing* encompasses all the various operations which can be applied to photographic or image data. These include, but are not limited to, image compression, image restoration, image enhancement, preprocessing, quantization, spatial filtering, and other image pattern recognition techniques.

99. *Interactive Image Processing* refers to the use of an operator or analyst at a console with a means of assessing, preprocessing, feature extracting, classifying, identifying and displaying the original imagery or the processed imagery for his subjective evaluation and further interactions.

100. *Pattern recognition* is concerned with, but not limited to, problems of: (1) pattern discrimination, (2) pattern classification, (3) feature selection, (4) pattern identification, (5) cluster identification, (6) feature extraction, (7) preprocessing, (8) filtering, (9) enhancement, (10) pattern segmentation, or (11) screening.

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## REFERENCES

- Afarani, M. K., Pattern Recognition Using Discrete Probability Distributions, Ph.D. Dissertation, University of Kansas, October, 1972.
- Anderson, J. R., E. E. Hardy, and J. T. Roach, "A Land-Use Classification System for Use with Remote-Sensor Data," Geological Survey Circular 671, U. S. Geological Survey, Washington, D. C., 1973.
- Cutrona, L. J., E. N. Leith, C. J. Palermo, and L. J. Porcello, "Optical Data Processing and Filtering Systems," IRE Transactions on Information Theory, vol. IT-6, no. 3, pp. 318-400, June, 1960.
- Darling, E. M., and R. D. Joseph, "Pattern Recognition from Satellite Altitudes," IEEE Transactions on Systems, Man, and Cybernetics, vol. SSC-4, pp. 38-47, March, 1968.
- Egbert, D., J. McCauley, and J. McNaughton, "Ground Pattern Analysis in the Great Plains," Semi-Annual ERTS A Investigation Report, Remote Sensing Laboratory, The University of Kansas Center for Research, Inc., Lawrence, Kansas, August, 1973.
- Fu, K. S., and J. M. Mendel, "Adaptive Learning and Pattern Recognition Systems," Academic Press, New York, 1972.
- Fukunaga, K., "Introduction to Statistical Pattern Recognition," Academic Press, New York, 1972.
- Goodman, J. W., "Introduction to Fourier Optics, McGraw-Hill Book Co., New York, 1968.
- Gramenopoulos, Nicholas, "Terrain Type Recognition Using ERTS-1 MSS Images," Symposium on Significant Results Obtained From the Earth Resources Technology Satellite, NASA SP-327, pp. 1229-1241, March, 1973.
- Haralick, R. M., et al, "Preliminary Report on Land Use Classification Using Texture Information in ERTS-A MSS Imagery," Report No. 2262-1, NASA Goddard Space Flight Center Contract NAS 5-21822, The University of Kansas Center for Research, Inc., Lawrence, Kansas January, 1973.
- Haralick, R. M., et al, "Texture-Tone Study with Application to Digitized Imagery," U. S. Army Engineer Topographic Laboratories Contract DAAK02-70-C-0388, CRES Technical Reports 182-2, and 182-3, The University of Kansas Center for Research, Inc., Lawrence, Kansas 1971, 1972.
- Haralick, R. M., "A Texture-Context Feature Extraction Algorithm for Remotely Sensed Imagery," Proceedings 1971 IEEE Decision and Control Conference, Gainesville, Florida, pp. 650-657, December, 15-17, 1971.
- Haralick, R. M., and D. E. Anderson, "Texture-Tone Study with Application to Digitized Imagery," CRES Technical Report 182-2, The University of Kansas Center for Research, Inc., Lawrence, Kansas, November, 1971.

- Haralick, R. M., K. Shanmugam, and I. Dinstein, "Texture Features for Image Classification," IEEE Transactions on Systems, Man, and Cybernetics, vol. SMC-3, pp. 610-621, November, 1973.
- Hardy, E. E., and J. R. Anderson, "A Land Use Classification System for Use with Remote-Sensor Data," Machine Processing of Remotely Sensed Data, (Conference Proceedings), Lafayette, Indiana, October, 1973.
- Horning, R. J., and J. A. Smith, "Application of Fourier Analysis to Multispectral/Spatial Recognition," Management and Utilization of Remote Sensing Data ASP Symposium, Sioux Falls, South Dakota, October, 1973.
- Kaizer, H., "A Quantification of Textures on Aerial Photographs," Boston University Research Laboratories, Technical Note 121, 1965, AD 69484.
- Kirvida, L., and G. Johnson, "Automatic Interpretation of ERTS Data for Forest Management," Symposium on Significant Results Obtained from the Earth Resources Technology Satellite, NASA SP-327, pp. 1076-1082, March, 1973.
- Kullback, S., Information Theory and Statistics, Wiley, 1957.
- Lendaris, G. G., and G. L. Stanley, "Diffraction-Pattern Sampling for Automatic Pattern Recognition," Proceedings of the IEEE, vol. 58, no. 2, pp. 198-216, February, 1970.
- Lewis, A. J., "Geomorphic Evaluation of Radar Imagery of Southeastern Panama and Northwestern Columbia," CRES Technical Report 133-18, The University of Kansas Center for Research, Inc., Lawrence, Kansas, February, 1971.
- MacDonald, H. C., "Geologic Evaluation of Radar Imagery from Darien Province, Panama," CRES Technical Report 133-6, The University of Kansas Center for Research, Inc., Lawrence, Kansas, July, 1969.
- Matherson, G., Elements Pour Une Theorie des Milieux Poreux, Masson, Paris, 1967.
- Miesel, W., "Computer Oriented Approaches to Pattern Recognition," Academic Press, New York, 1972.
- O'Neill, E., "Spatial Filtering in Optics," IRE Transactions on Information Theory, vol. IT-2, no. 2, pp. 56-65, June, 1956.
- Preston, K., Coherent Optical Computers, McGraw-Hill Book Co., New York, 1972.
- Read, J. S., and S. N. Jayaramamurthy, "Automatic Generation of Texture Feature Detectors," IEEE Transactions on Computers, vol. C-21, no. 7, pp. 803-812, July, 1972.
- Rosenfeld, A., and M. Thurston, "Edge and Curve Detection for Visual Scene Analysis," IEEE Transactions on Computers, vol. C-20, no. 5, pp. 562-569, May, 1971.

- Rosenfeld, A., and E. Troy, "Visual Textural Analysis," University of Maryland Computer Science Center, TR-70-116, June, 1970.
- Serra, J., and G. Verchery, "Mathematical Morphology Applied to Fibre Composite Materials," Film Science and Technology, vol. 6, pp. 141-158, 1973.
- Shulman, A. R., Optical Data Processing, John Wiley & Sons, Inc., New York, 1970.
- Sutton, R., and E. Hall, "Texture Measures for Automatic Classification of Pulmonary Disease," IEEE Transactions on Computers, vol. C-21, no. 7, pp. 667-676, July, 1972.
- Swanlund, G., "Honeywell's Automatic Tree Species Classifier," Honeywell Systems and Research Division, Report 9D-G-24, December 31, 1969.
- Yaglom, A. M., Theory of Stationary Random Functions, Prentice-Hall, Inc., New Jersey, 1962.