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TECHNICAL MEMORANDUM EVALUATION OF THREE-CATEGORY CLASSIFICATION

Ву

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(E80-10058) LARGE AREA CROP INVENTORY EXPERIMENT (LACIE). EVALUATION OF THREE-CATEGORY CLASSIFICATION (Lockheed Electronics Co.) 45 p HC A03/MF A01

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PREFACE

Documented in this report are the results of a study to evaluate the Large Area Crop Inventory Experiment clustering and classification procedures in terms of variance of the proportion estimates and the probabilities of correct classification for three categories. The categories of interest were corn, soybeans, and other.

Timely preparation of the data and experiment design for this study would not have been possible without the aid of several coworkers. K. Lennington and D. Register wrote the initial experiment design. R. Abotteen and J. Johnson helped to verify the ground-truth labels and to prepare the initial machine processing runs. Their assistance with this study was greatly appreciated.

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ACRONYMS

ANOVA analysis of variance

LACIE Large Area Crop Inventory Experiment

Landsat land satellite

ME machine estimate

P1 Procedure 1

PCC probability of correct classification

PCC1 probability of correct classification for type 1 dots

PCC2 probability of correct classification for type 2 dots

pixels picture elements

R reduction coefficient

SAE stratified areal estimate

TY Transition year

1. INTRODUCTION

In transition from the Large Area Crop Inventory Experiment (IACIE) to the LACIE Transition Project (FY79), the basic components of Procedure 1 (P1) required investigation. Pl, as used in LACIE, was a two-category procedure estimating wheat and nonwheat. In the mixed wheat states, a threecategory classification was used to estimate winter wheat, spring wheat, and other in LACIE Phase III, but no investigation of the appropriate number of type I dots was made. This study was initiated to test a three-category classifier using corn, soybeans, and other as categories to examine the appropriate number of type 1 dots. Since a machine estimate (ME) and a stratified areal estimate (SAE) were produced by both nearest-neighbor clustering and maximum-likelihood classification in a standard Pl run, all four estimates were compared to ground-truth proportions. Each of these four proportion estimates were also analyzed in terms of the variance of the estimates and the variance of the 'orresponding probabilities of correct classification (PCC). The reduction coefficients, R-values, were calculated for all processing runs and were compared to previous two-category calculations.

2. EXPERIMENT OBJECTIVES AND DESCRIPTION

2.1 OBJECTIVES

The experiment was designed to test the three-category classifier using corn, soybeans, and other as the categories. The objectives of the experiment were:

- a. To examine a three-category classifier proportion estimate in terms of the number of type 1 dots used
- b. To examine a three-category classifier in terms of the variance of the estimate
- c. To examine the evaluation criterion (the PCC) in terms of its variance

2.2 GENERAL DESCRIPTION

The experiment was planned to include the processing of 12 test segments using varying numbers of starting dots (type 1) and 105 bias correction dots (type 2). Of these 12 segments, 6 were obtained from the corn and soybean allocation and 6 from the LACIE Transition Year (TY) allocation. Detailed information about these data is described in table 2-1.

The crops chosen for both the six segments obtained from the corn and soybean allocation and the six segments obtained from the LACIE TY allocation were the major crops in the segment. The primary purpose of the test was to process corn and soybeans; however, if one of these crops were not adequately represented in a segment, another major crop was chosen to replace it as a crop of interest.

The number of type 1 dots (sets of 30, 45, and 60 dots) was varied in order to examine the effect of the number of dots used in a three-category classifier. To estimate the variance of the proportion estimates and the PCC, three independent sets of dots were selected from the 209 grid intersections for a fixed number of type 1 dots. Thus, a total of 108 processing runs was possible. To make the initial type 1 dot selections for a segment, three independent sets of 60 dots were randomly selected in the usual manner of skipping all border

TABLE 2-1 .- DATA SET

	Segment	Naguiaitian wand	Maday ayang				
Number	Location	Acquisition used	Major crops				
146	Kentucky	8180, 8198, 8234, 8270	Corn (C), Soybeans (X)				
185	Minnesota	8169, 8197, 8205, 8224	Corn (C), Soybeans (X), Spring wheat, Sunflowers				
804	Iowa	8229, 8247, 8274, 8292	Corn (C), Soybeans (X) Oats				
812	Mississippi	8199, 8235, 8280, 8289	Soybeans (X), Cotton (K), Rice				
824	Illinois	8163, 8235, 8271, 8307	Corn (C), Soybeans (X)				
883	I owa	8186, 8213, 8222, 8293	Corn (C), Soybeans (X)				
1075	Nebraska	8133, 8206, 8259, 8296	Corn (C), Alfalfa (A)				
1253	Ok1ahoma	8165, 8184, 8274, 8291	Soybeans (X), Alfalfa (A)				
1341	Kansas	8113, 8167, 8186, 8293	Corn (C), Soybeans (X) Sorghum				
1502	Colorado	8138, 8246, 8282, 8300	Corn (C), Sugar beets (Y) Winter wheat, Alfalfa				
1572	Nebraska	8153, 8206, 8279, 8296	Corn (C), Pasture (P)				
1591	Nebraska	8134, 8241, 8259, 8278	Corn (C), Sorghum (E)				

or edge picture elements (pixels) described in reference 1. Next, 15 dots were randomly deleted to produce three sets of 45 type 1 dots. And again, 15 dots were randomly deleted for three sets of 30 type 1 dots. For both of these random deletions, each category was guaranteed to have at least one type 1 dot, thus restricting the deletion process.

Each processing consisted of a three-category version of the standard P1 clustering and classification. Proportion and PCC estimates were obtained from the automatically labeled clusters and from the maximum-likelihood classifier output. The SAE were also calculated for each ME using a set of 105 type 2 dots. For the three replications, using a fixed number of starting dots (30, 45, or 60), the type 2 dots were selected independently where the overlap (between sets of type 2 dots only) occurred from necessity. The abundance of border and edge pixels in the type 1 selections prohibited a third set of dots for three of the twelve test segments. This caused the total number of processing runs to be decreased to 99.

2.3 EXPERIMENT DESIGN

An analysis of variance (ANOVA) was planned for each set of starting dots (set of 60, set of 45, and set of 30) to determine differences between the proportion estimation procedures. The signed difference between each proportion estimate and the ground-truth estimate was used for the response variable. The linear model for the three analyses was as follows:

$$Y_{ijkm} = \mu + \sigma_i + \tau_j + C_k + \varepsilon_{ijkm}$$

where

 μ = the overall mean of the observations

 σ_i = the segment effect (i = 1, 2, ..., 12)

 τ_j = the treatment or procedure effect (j = 1, 2, 3, 4)

 C_k = the crop effect (k = 1, 2)

 ε_{ijkm} = the random error for each observation (m represents the repetitions performed for each observation and is a function of i, j, and k.)

 Y_{ijkm} = the response variable

An ANOVA was also planned for each proportion estimation procedure to determine differences between the number of starting dots used (60, 45, and 30). The response variable was again the signed difference between the proportion estimate and ground truth. The linear model for these four ANOVAs was as follows:

$$Y_{ijkm} = \mu + \sigma_i + \tau_j + C_k + \epsilon_{ijkm}$$

where

 τ_j = the treatment effect representing the number of starting dots used (j = 1, 2, 3)

 μ = the overall mean of the observations

 σ_i = the segment effect (i = 1, 2, ..., 12)

 C_k = the crop effect (k = 1, 2)

 ε_{ijkm} = the random error for each observation (m represents the repetitions performed for each observation and as a function of i, j, or k.)

A general linear model ANOVA program was used to generate the ANOVA tables (ref. 2).

To examine the variability in the performance of P1, estimates of the variance of the proportion estimates and the variance of the PCC estimates were to be computed. The variances were estimated by pooling the within-segment variances over each segment for each case of 30, 45, and 60 type 1 dots. These variances were then pooled over all segments for each case of 30, 45, and 60 type 1 dots. The equations for computing these variance estimates are as follows.

$$Var [X] = \frac{\sum_{j=1}^{N} \sum_{j=1}^{M_{j}} (x_{ij} - \overline{x}_{j})^{2}}{2N}$$

where

 X_{ij} * the variable, proportion estimate, or PCC as measured for the 1th sampling and jth segment

 \overline{X}_{j} = the average value of $X_{j,j}$ for the jth segment

N = total number of segments and range of j

M_j = total number of samplings which are dependent upon j and are in the range of i

Separate comparisons were planned for the variance of the proportion estimates and the variance of the PCC estimates. In each case, ratios between the variances for the estimates of the set of 45 type 1 dots and the set of 60 type 1 dots and between the variances for the estimates for the set of 30 and the set of 60 type 1 dots were to be calculated. These ratios were approximately distributed as F-statistics and, therefore, may be tested for statistically significant departures from unity. Statistical tables indicated that ratios with a value of approximately two or larger were significant at the 5-percent level if a total of 12 segments was used.

3. PROCEDURAL DESCRIPTION

This study was performed using ground-truth labels that were manually verified with an annotated aerial photograph and registered grid overlay. The grid overlay corresponds to the grid intersections on the land satellite (Landsat) film products. Border (spectrally mixed pixels) and edge (spatially misregistered pixels from acquisition to acquisition) were also identified and documented at this time since these types of pixels are not used as type 1 dots.

Standard P1 processing was performed. The type 1 dots started the nearest-neighbor clustering algorithm (ref. 3) with the following parameters.

a. CO	Ħ	60
-------	---	----

- b. Percent = 0
- c. SEP = 1
- d. STDMAX = 20
- e. DLMIN = 0
- f. R2 = 8191
- g. NMIN2 = 18
- h. ITMAX = 0
- i P of N = 1
- j. SC Seq. = S
- k. Distance measure = L2 (Euclidean)

The NMIN2 parameter was changed from the standard value of 100 to 18 in order to prevent the deletion of small clusters.

The clusters were automatically labeled by the closest type 1 dot using an L2 distance criterion. The cluster statistics were then used in a maximum-likelihood classifier to classify the segment. Output reports included

cluster proportion estimates, classification proportion estimates, their corresponding SAEs, type 1 PCCs, and type 2 PCCs.

Initially, the three-category version of P1 was run using each set of 60 type 1 dots for each segment. Following the completion of these runs, 15 type 1 dots were deleted at random from each set of type 1 dots and the processing was repeated using the 45 remaining dots. Finally, 15 more type 1 dots were deleted at random, and the 30 remaining dots were used to make the final runs.

4. RESULTS

The estimates obtained from the study are shown in tables 4-la, 4-lb, and 4-lc for the sets of 60, 45, and 30 type l dots, respectively. The Cl and C2 are the two categories of interest that were processed with other (N). The ground-truth estimates of these categories are 400-random-dot counts, taken from annotated aircraft photography because digitized ground-truth maps were not available. The ME, SAE, type l dot PCC (PCCl), and type 2 dot PCC (PCC2) are shown for both cluster and classification results. The SAEs were computed on a category level for both MEs. Note that the PCC values are computed for the ME only.

The raw proportion estimates were differenced with the ground truth before analysis, and these signed differences appear in the appendix, table A-la, A-lb, and A-lc for 60, 45, and 30 type 1 (14), respectively.

The first set of ANOVA tests was performed on the signed differences between proportions and ground truth for each different number of type 1 dots: 60, 45, and 30. This was to determine if any significant differences existed between the four methods of achieving a proportion estimate. These ANOVA tests appear in table 4-2a, 4-2b, and 4-2c. For each separate set of starting dots, no significant differences were found between the proportion estimates.

The ANOVA tests were also performed to detect differences between 30, 45, and 60 starting dots, based on the signed differences between the proportions and ground truth. These ANOVA tests appear in tables 4-3a, 4-3b, 4-3c, and 4-3d. No significant differences were found between numbers of starting dots (60, 45, and 30) for each proportion estimation technique: machine clustering, SAE clustering, machine classification, and SAE classification.

The variance of the proportion estimates on a per segment basis appear in the appendix, tables A-2a, A-2b, and A-2c for 60, 45, and 30 starting dots, respectively. Comparisons were more readily made when these variances were pooled over all the test segments for each number of starting dots, as in

TABLE 4-1.- RAW DATA ESTIMATES FOR 60, 45, AND 30 TYPE 1 DOTS

(a) 60 type 1 dots

								CT	uster					Sia	1551Fy		
3	Segment		Sround-truth data					ed areal	PCC1	PCCF	Machine estimate		Stratified areal estimate		8C51	2002	
Yimter	Location	Crop 1	Percent	Erop 2	Fercent	Sesp I	Crsp Z	Crop 1	Crop 2			Sess 7	2rc2 2	Sess I	Srop 2		
1075	Nebreska	С	23	ř.	2	15 25 25	4 8 6	23 36 34	14 71 12	190.0 195 195	72_4 75,2 77,1	25 25 25	9.819	25 33	14 11 12	136.0 136.0	69.5 72.1 75.8
1341	Kansas	c	40	1	14	45 41 42	15 18 20	35 39 36	17 16 17	90.3 95.3 100	55.2 56.2 55.2	£7 £7 42	15 17 20	25 27 37	18 14 7	96.3 96.3 100	59.2 57.1 55.9
1551	Mebraska	С	15	£	7	15 10 14	5 3 7	19 12 15	4 5 6	98.3 100 98.3	83.3 79.6 75.2	15 15	3 5	:E 17 16	3 5 7	96.3 190 190	23.2 79.6 81.6
146	Kentucky	С	17	Į.	46	75 17	u u	20 16	45	38.3 100	73.8 73.7	75 76	44 46	25 17	45 45	98.3 98.3	79.8 72.7
125	Minnesota	٤	s	x	8	4 3	£ 2	7 & 8	10 6	100 100	87.6 85.7 82.9	3	5 2 4	7 5 8	:57	130 130 130	87.3 85.€ 83.7
804	lon	c	45	1	z	47 45 43	2E 21 21	40 44 41	у 23 27	130 130	82.0 79.5 82.9	47 45 43	25 27 27	47 42 41	33 29 27	136 136 136	75.G 76.8 82.7
212	Mississippi	1	42	K	7	49 50 51	m et m	50 52 42	5 6 7	196 98.3 196	80.0 81.5 72.1	48 49 50	5 21 45	50 53 42	<u> </u>	190 90.3 190	87.8 84.5 78.1
824	Illinois	c	5 2	x	41	55 52	41 43	52 45	27 44	120 95.7	75.0 81.5	55 52	43 43	52 48	3£ 42	130 \$6.6	77.5 82.5
823	lowa	c	37	x	33	32 41	3C 31	40 35	34 33	96.3 96.3	63.5 69.2	33 33	31 21	41 36	33 33	%.6 %.7	65.6 5.17
1253	Oklahom	x	23.	A	3	33 33	1	29 33 25	2 2 2	%.7 %.3 100	87.7 83.8 81.0	32 32 32	1	31 33 27	3 2	96.6 190 700	92.3 85.6 85.1
1502	Colorado	t	11	Y	4	9 3 10	1 1 5	12 8	2 6 7	100 96.3 100	83.8 86.7 86.7	9 2 10	1 5	12 3	5 7	96.3 100 100	86.4 86.3 86.5
1572	Nebraska	c	14	P	64	5 14 17	76 75 69	14 14 15	59 54 63	95.3 95.0 95.3	72_4 75.5 74.3	5 14 6	76 75 70	16 74 76	59 54 55	笑.7 笑.7 笑.3	73.2 71.4 73.3

Symbol definitions:

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans

1 * sugar beets



TABLE 4-1.— Continued.

(b) 45 type 1 dots

								C)	uster					Classify			
	egment	Ground-truth data				Hechine S estimate		Stratified areal estimate		PCC2	Machine estimate		Stratified areal estimate		PCC1	PCC 2	
Number	Location	Crop i	Percent	Crop 2	Percent	Crap 1	Crop 2	Crop I	Crop 2			Crop 1	Crop 2	Crop 1	Crop 2		{
1075	Nebraska	C	29	A	8	17 25 26	4 7 6	28 33 33	14 12 10	100.0 100 100	72.4 73.1 78.1	17 24 25	4 7 E	27 34 33	14 12 10	100.0 100 100	70.9 76.7 79.8
1341	Kansas	{ с 1	40	x	14	47 37 44	17 21 23	35 39 38	17 15 15	97.8 100 100	53.3 53.3 50.5	47 37 44	17 20 22	36 39 39	17 16 16	97.8 100 100	55.9 58.1 59.2
1591	Kebraska	c	15	E	7	17 10 14	4 1 8	17 17 16	3 5 7	93.3 100 97.7	79.4 79.0 78.8	17 9 14	4 1 7	17 18 17	3 5 6	88.9 100 100	80.4 81.0 80.4
146	Kentucky	С	17	x	45	12 17	50 45	18 18	47 44	97.8 100	81.0 82.0	12 16	50 45	19 17	45 43	97.8 100	81.9 83.1
185	Rinnesota	c	В	x	8	4 3 1	7 2 8	5 5 8	6 11 7	95.5 100 100	84.8 85.7 80.0	4 3 1	7 2 8	5 5 8	6 11 7	97.7 100 100	84.2 85.4 81.6
804	lom	c	46	x	29	44 45 45	25 27 28	40 42 41	34 30 28	190 100 100	79.0 81.0 81.9	44 45 44	26 27 28	40 42 41	34 29 25	100 100 97.8	78.2 80.8 81.7
812	Mississippi	x	48	K	7	48 48 50	7 3 6	50 49 43	5 4 7	95.6 100 100	80.0 83.8 78.1	48 48 49	2 3 6	49 53 43	6 5 7	95.5 100 97.7	61.7 50.8 78.8
824	illinois	C	52	x	41	54 51	42 43	54 47	34 44	97.8 100	77.5 81.6	54 51	42 43	53 49	35 43	97,8 100	78.8 82.3
883	lom	c	37	x	33	35 38	31 30	41 38	34 33	97.8 100	66.7 70.9	35 38	32 30	42 38	32 33	95.6 97.8	65.3 72.9
1253	OkTahoma	×	33	٨	3	35 27 33	1	29 33 28	2 3 2	97.8 95.6 100	86.8 81.9 84.2	35 27 33	1	32 33 28	3 3 1	100 95.6 100	86.5 82.5 86.7
1502	Colorado	c	11	٧	4	<i>1</i> 3 11	17	6 12 8	7 5 8	100 100 100	82.9 87.6 85.7	7 3 10	2 1 7	6 12 9	6 5 7	100 100 100	82.4 87.4 85.6
1572	Rebraska	c	14	P	64	6 14 5	81 75 72	14 14 14	93 54 65	95,6 95,6 93,3	67.6 70.5 69. 5	6 14 6	81 75 72	14 12 14	59 56 64	95,6 93,3 93,2	68.6 72.8 70.5

Symbol definitions:

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans



TABLE 4-1.— Concluded.

(c) 30 type 1 dots

								CI	uster					Cia	ssify		
	itgment	Ground-truth data			Hachine Stratified areal estimate PEC: PCC2			PCC2	Machi estin			ed areal	PCC?	PCCZ			
Number	Location	Crop 1	Percent	Crop 2	Percent	Crap 1	Crop 2	Crop 1	Crop 2			Crop 1	Crop 2	Crop 1	Crop 2		
1075	Nebraska	ε	23	Å	S	17 26 23	2 15 6	29 33 32	13 10 8	160.0 100 100	67.6 74.0 74.3	17 26 22	1 13 6	29 33 31	13 12 2	195.5 196 196	62.3 76.7 79.6
1341	Kansas	С	40	x	14	53 31 46	18 29 31	35 43 32	18 14 15	90.5 95.7 100	55.2 52.4 45.7	52 31 45	18 28 31	37 44 37	17 16 16	90.0 195 96.6	55.3 55.2 47.6
1591	Rebraska	С	15	E	7	15 10 10	3 1 5	15 17 16	3 5 7	96.7 100 100	75.5 73.1 75.6	15 15 11	2 1 4	17 16 16	3 5 7	95.7 199 199	80.0 79.6 75.7
146	Kentucky	c	17	X	46	14 15	53 45	20 17	47 42	80.0 100	71.4 75.2	14 16	54 45	21 18	45 43	86.7 100	£9.9 1.£5
185	Minnesota	E.	8	X	8	0 2 5	5 3 3	5 3 8	6 9 6	96.6 100 96.7	23.3 23.2 76.2	0 2 5	5 3 3	เพล	5 10 6	96.5 100 89.7	85.1 84.0 75.2
804	i Owek	C	46	x	29	44 49 43	2£ 27 26	41 42 42	35 29 22	100 100 100	76.2 79.0 81.0	45 42 47	27 27 27	43 43 41	34 29 22	100 100 100	77.7 79.6 21.6
212	Mississippi	X	48	ĸ	7	43 50 55	5 3 7	49 47 43	6 8	190 190 93.3	78.1 82.9 78.1	42 50 54	5 4 7	49 52 42	6 7 E	196 100 56.8	81.7 85.4 78.8
224	Illinois	c	52	x	41	57 51	32 41	50 47	37 42	100 100	77.5 75.5	57 50	29 42	49 49	36 43	160 169	75.0 72.9
223	Iona	С	37	x	33	42 44	32 32	39 34	34 37	100 100	64.6 72.7	41 43	32 32	47 35	35 34	95.7 100	66.G 72.1
1253	Oki a homa	x	33	Ā	3	32 24 33	3 1	29 33 22	2 3 2	95.7 100 100	26.2 20.0 84.2	33 24 33	1 3 T	31 33 28	3 4 1	169 106 100	85.7 81.7 86.1
1502	Colorado	С	7.1	Y	4	12 12	2 1 3	6 12 9	7 4 7	100 100 100	82.9 87.5 89.5	4 4 12	2 1 3	7 12 9	6 5 7	150 150 150	62.8 67.5 89.9
1572	Kebraska	С	14	₽	64	7 3 6	22 61 77	15 12 14	52 57 66	83.3 83.8 93.3	70.5 62.6 62.6	7 9 6	27 21 72	15 11 13	59 52 66	83.3 96.5 90.0	72.4 69.9 70.5

Symbol definitions:

Crop codes

A = alfalfa

C * com

E = sorghum

K = cotton

P = pasture

X = soybeans

Y = sugar beets

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TABLE 4-2.— PROPORTION ESTIMATES ANOVA USING 60, 45, AND 30 STARTING DOTS

Source of variation	Degrees of freedom	Sum of squares	Mean square	F-value				
(a)	Using 60 sta	Using 60 starting dots						
Mean, segment, and crop effect	13	505.31						
Estimation procedure effect	3	32.59	10.86	0.85 ^a				
Error	248	3149.10	12.70					
Total	264	3687.00						
(b)	Using 45 sta	arting dots						
Mean, segment, and crop effect	13	697.02						
Estimation procedure effect	3	10.80	3,60	0.25 ^a				
Error	248	3558.17	14.34					
Total	264	4266.00						
(c)	Using 30 sta	arting dots						
Mean, segment, and crop effect	13	1223.56						
Estimation procedure effect	3	1,85	0.61	0.03 ^a				
Error	248	5564.61	22.44					
Total	264	6790.00						

 $^{^{\}rm d}Indicates$ nonsignificance at the α = 5-percent level.

TABLE 4-3.— STARTING DOTS ANOVA USING MACHINE CLUSTERING PROPORTIONS, SAE CLUSTERING PROPORTIONS, MACHINE CLASSIFICATION PROPORTIONS, AND SAE CLASSIFICATION PROPORTIONS

	····	1		
Source of variation	Degrees of freedom	Sum of squares	Mean square	F-value
(a) Ma	chine cluster	ing proport	ions	
Mean, segment, and crop effect	13	1918.96		
Starting dot effect	2	14.92	7.46	0.38 ^a
Error	183	3569.12	19.50	
Total	198	5503.00		
(b)	SAE clusteri	ng proportio	ons	
Mean, segment, and crop effect	13	461.60		
Starting dot effect	2	3.67	1.83	0.22 ^a
Error	183	1550.73	8,47	
Total	198	2016.00		
(c) Mac	hine classific	cation propo	ortions	
Mean, segment, and crop effect	13	1968.49		
Starting dot effect	2	17.04	8.52	0.46 ^a
Error	183	3394,47	18.55	
Total	198	5380.00		
(d) S	AE classificat	ion proport	tions	
Mean, segment, and crop effect	13	487.19		
Starting dot effect	2	0.39	0.20	0.03 ^a
Error	183	1356.42	7.41	
Total	198	1844.00		

 $^{^{}a}\text{No}$ significant difference was found at the α = 5-percent level.

table 4-4. The variances of the MEs, both clustering and classification, were significantly decreased by increasing the number of starting dots from 30 to 45 for both categories of interest. These variances were again decreased, but not significantly, by increasing the numbers of starting dots from 45 to 60. The variances of the SAE estimates, both clustering and classification, did not significantly differ for any change in the number of starting dots. For 60 starting dots, no significant differences were found between any of the proportion estimation techniques for either crop.

The variances of the PCCs on a per segment basis appear in the appendix, tables A-3a, A-3b, and A-3c for 60, 45, and 30 starting dots, respectively. These variances are pooled over all test segments for each number of starting dots and appear in table 4-5. The variance of the PCCl for clustering and classification was significantly decreased when the number of starting dots changed from 30 to 45 and from 45 to 60. The variance of the PCC2 for clustering and classification was significantly decreased when the number of starting dots changed from 30 to 45. When the number of starting dots changed from 45 to 60, the variance of PCC2 increased for both clustering and classification, but the increase was not significant. For 60 starting dots, the variance of PCCl was significantly different from the variance of PCC2 for both clustering and classification. This significance can be attributed to the difference between training and test data.

The reduction coefficient, R, has been presented as a method of observing how much the machine classification reduces the variance of the SAE proportion estimation (ref. 3) in comparison with the variance of a simple random sample estimate. In the computation of the R-values, the omission and commission rates are computed by comparing the machine labels of the type 2 dots to the ground truth. The sampling error can be computed using the ground truth proportions or the labeling proportion from the type 2 dots denoted herein as π . The R-values were computed for both clustering and classification for all of the three-category runs. The omission and commission rates and the R-values are presented, for both ground truth and labeling proportions, in the appendix, tables A-4a, A-4b, and A-4c. The R-values were then averaged over the 33 runs for a particular number of starting dots (60, 45, or 30) and using

TABLE 4-4.- VARIANCES OF THE PROPORTION ESTIMATES

Number of	Cnan	Clus	ter	Classify			
starting dots	Crop	Machine	Bias	Machine	Bias		
30	1	23.389	5.562	22.437	6.549		
30	2	14.021	6.312	12.875	4.597		
AF	1	11.285	5.181	10.625	6.111		
45	2	6.479	7.431	6.181	5.076		
60	1	7.715	8.194	8.007	6.542		
60	2	3.493	5.125	3.083	5.264		

^aStatistical significance at the 5-percent level is found whenever the ratio of variances is at least two.

TABLE 4-5.- VARIANCES OF THE PCC

Number of	Clu	ster	Classify			
starting dots	PCC1	PCC2	PCC1	PCC2		
30	16.751	10.992	12.782	14.446		
45	3,053	4.590	5,254	5.484		
60	1.138	5.725	0.954	7.408		

both the ground-truth and the labeling proportion. The average R-values and their corresponding variances are presented in table 4-6. The averaged R-values were only slightly higher than those for the two-category P1 runs (0.718 and 0.714) as documented in reference 3.

TABLE 4-6.— THE R-VALUES AVERAGED OVER 33 THREE-CATEGORY RUNS

Number of		CT	luster	Classify		
starting dots	π	R-mean	R-variance	R-mean	R-variance	
60	Ground truth	0,738	0.024	0.731	0.034	
60	Labeling proportion	.739	.025	.729	.032	
45	Ground truth	.736	.032	.730	.029	
45	Labeling proportion	.735	.032	.729	.030	
30	Ground truth	.775	.029	.746	.034	
30	Labeling proportion	.774	.029	.746	.034	

2/2

5. CONCLUSIONS AND RECOMMENDATIONS

Upon examination of both ME and SAE proportion estimates produced by clustering and classification, no significant differences were found between the proportion estimates and ground-truth estimates. Since this was the case in previous two-category studies (ref. 4), it is not considered unusual in the three-category case, but instead, indicates that conclusions should be made on the basis of the consistency or variance of the estimates as well as the accuracy.

When testing the variances of the ME proportion estimates, a significant reduction in the variances was found when the number of starting dots was increased from 30 to 45. The variances were again reduced, although not significantly, when the number of starting dots was increased from 45 to 60. From these results, 60 starting dots are recommended for a three-category classifier.

When examining the variances of the estimates for the four estimation procedures (using 60 starting dots), no significant differences were found between procedures. Thus, only the machine clustering may be used to produce an estimate and the SAE computations and maximum-likelihood classification can be deleted. This will allow two advantages over PI: (1) using only clustering will eliminate the additional machine time required by classification, and (2) deleting the SAE will minimize the analyst-labeling-time required because only type 1 dot labeling will be necessary.

The variance of the PCC1 was significantly lower for 60 starting dots than for either the set of 45 or 30 dots. Since the type 1 dots are the training data, an increase in the training sample size is expected to produce significant decreases in the variance of the PCC1. For the PCC2, a significant reduction in the variance was observed when the number of starting dots was increased from 30 to 45. No significant differences were observed when the number of starting dots was increased from 45 to 60. Thus, the variance of

PCC1 decreased when the number of starting dots increased up to 60, and the variance of PCC2 decreased when the umber of starting dots increased to 45 and then statistically stabilized. This further reinforces the choice of 60 starting dots.

The efficiency of P1 in reducing the variance of a proportion estimate obtained from SAE has been presented in reference 3. In this experiment, virtually no difference existed between the R-values, regardless of the number of starting dots used or the proportion estimation procedure. There were no cases where the R-value was lower for clustering than for classification. This would indicate that classification was better than clustering, but the differences between the R-values were consistently very small. As in the two-category case, these R-values indicate that not much is being gained by classification or clustering over a simple random sample. Since these R-values are the product of the best possible labeling of dots (ground truth), an improved procedure to P1 seems desirable to improve the cost-effectiveness of this machine processing.

To summarize, the recommendations resulting from this study are as follows:

- a. A set of 60 starting dots should be used in a three-category classifier.
- b. The ME produced by nearest-neighbor clustering is an adequate estimator.
- c. More study is needed in the area of an alternative for Pl.

6. REFERENCES

- 1. LACIE Transition Project (FY79) Detailed Analysis Procedures. LACIE-LACIE-00724, JSC-13756, March 1979.
- 2. Graybill, Franklin A.: Theory and Application of the Linear Model. Duxbury Press, North Scituate, Mass., 1976, pp. 247-252.
- 3. Havens, K. A.: Further Evaluation of Procedure 1 Secondary Error Analysis. LEC-13180, May 1979.
- 4. Havens, K. A.: Secondary Error Analysis: The Evaluation of Analyst Dot Labeling. LEC-12380, September 1978.

APPENDIX

APPENDIX

Tables A-1 through A-5, included in the appendix, are supplemental material referred to in section 4 of this document.

TABLE A-1.— DIFFERENCES BETWEEN PROPORTIONS AND GROUND TRUTH USING 60, 45, AND 30 STARTING DOTS

(a) Using 60 starting dots

[C)u	ster			Clas	sify	
	Segment	Crop 1	Crop 2	Machine (estimate	Stratific esti	ed areal mate	Machine estimate		Stratified areal estimate	
Number	Location			Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2
1075	Nebraska	С	A	-13 -3 -3	-4 0 -2	-1 7 5	6 3 4	-13 -4 -3	-5 0 -2	-1 7 4	6 3 4
1341	Kansas	С	X	6 1 2	1 4 6	-5 -1 -4	3 2 3	7 1 2	1 3 6	-5 -3 -3	4 2 3
1591	Nebraska	С	E	0 -5 -1	-2 -4 0	4 3 0	-3 -2 -1	1 -5 -1	-3 -4 -2	1 2 1	-4 -2 0
146	Kentucky	С	X	-2 0	-2 0	3 -1	0 -2	-2 -1	-2 0	3 0	0 -1
185	Minnesota	С	X	-4 -5 -7	-2 -6 -4	-1 -2 0	-2 2 -2	-4 -5 -7	-2 -6 -4	-1 -2 0	-2 2 -1
804	Iowa	С	х	-1 -3	-3 -2 -2	-6 -2 -5	5 0 -2	2 -1 -3	-3 -2 -2	-5 -4 -5	4 0 -2
812	Mississippi	x	к	1 2 3	-2 -4 -2	2 4 -6	-2 -1 0	0 1 2	-2 -4 -2	2 5 -6	-1 -1 0
824	Illinois	С	X	3 0	0 2	0 -6	-4 3	3 0	0 2	0 -4	-5 1
883	Iowa	С	X	1 3	-3 -2	.]	1 0	1 2	-2 -2	4 1	1
1253	Oklahoma	х	Α	-1 0 -1	-2 -2 -2	-4 0 -7	-1 0 -1	-1 0 -1	-2 -2 -2	-2 0 -6	-2 0 -1
1502	Colorado	С	Y	-2 -8 -1	-3 -3 1	-5 1 -3	4 0 3	-2 -8 -1	+3 -3 1	-5 1 -2	2 1 3
1572	Nebraska	С	Р	-8 0 -7	12 11 5	0 0 1	-5 -10 -1	-8 0 -8	12 11 6	0 0 0	-5 -10 1

Symbol definitions:

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans



TABLE A-1.— Continued.

(b) Using 45 starting dots

					Clu	ster			Clas	sify	
s	egment	Crop 1	Crop 2	Machine	estimat e	Stratific estin	ed areal mate	Machine	estimate	Stratif; esti	
Number	Location			Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	Crop l	Crop 2
1075	Nebraska	С	A	-12 -4 -3	-4 -1 -2	-1 -4 -4	6 4 2	-12 -5 -3	-4 -1 -2	-2 5 4	6 4 2
1341	Kansas	С	X	7 -3 4	3 7 9	-5 -1 -2	3 1 1	7 -3 4	3 6 8	-4 -1 -1	3 2 2
1591	Nebraska	C	E	2 -5 -1	-3 -6 1	2 2 1	-4 -2 0	2 -6 -1	-3 -6 0	2 3 2	-4 -2 -1
146	Kentucky	С	Х	-5 0	4 -1	1	1 -2	-5 -1	-1	2 0	-1 -3
185	Minnesota	C	X	-4 -5 -7	-1 -6 0	-3 -3 0	-2 3 -1	-4 -5 -7	-1 -6 0	-3 -3 0	-2 3 -1
804	Iowa	С	x	-2 -1 -1	-3 -2 -1	-6 -4 -5	5 1 -1	-2 -1 -2	-3 -2 -1	-6 -4 +5	5 0 -1
812	Mississippi	X	К	0 0 2	0 -4 -1	2 1 -5	-2 -3 0	0 0 1	1 -4 -1	1 5 -5	-1 -2 0
824	Ill inois	С	Х	2 -1	1 2	2 -5	-7 3	2 -1	1 2	1 -3	-6 2
883	Iowa	С	х	-2 1	-2 -3	1	0	-2 1	-1 -3	5 1	-1 0
1253	Oklahoma	x	Α	2 -6 0	-2 -2 -2	-4 0 -5	-1 0 -1	2 -6 0	-2 -2 -2	-1 0 -5	0 0 -2
1502	Col orado	С	Y	-4 -8 0	-2 -3 3	-5 1 -3	3 1 4	-4 -8 -1	-2 -3 3	-5 1 -2	2 1 3
1572	Nebraska	С	Р	-8 0 -9	17 11 8	000	-6 -10 1	-8 0 -8	17 11 8	0 -2 0	-5 -8 0

Symbol definitions:

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans

TABLE A-1.- Concluded.

(c) Using 30 starting dots

					Clu	ster			Clas	sify	
S	iegment	Crop 1	Crop 2	Machine	estimat e	Stratific estin	ed areal mate	Machine	estimate	Stratific esti	ed areal
Number	Location			Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crap 2	Crop 1	Crop 2
1075	Nebraska	С	A	-12 -13 -1	-6 2 -2	0 4 3	5 % 0	-12 -3 -1	-7 2 -2	0 4 2	5 4 0
1341	Kansas	C	х	13 -9 6	4 15 17	-5 3 -2	4 0 1	12 -9 6	4 14 17	-3 4 -3	3 0 2
1591	Nebraska	С	ε	0 -5 -5	-4 -6 -2	0 2 1	-4 -2 0	0 -5 -4	-5 -6 -3	2 1 1	-4 -2 0
146	Kentucky	С	X	-3 1	7 -2	3 0	-4	-3 0	-2	4 2	-1 -4
185	Hinnesota	С	X	-8 -6 -3	-3 -5 -5	-3 -5 0	-2 1 -2	-8 -6 -3	-3 -5 -5	-3 -5 0	-2 2 -2
804	lowa	С	х	-2 3 2	-3 -2 -3	-5 -4 -4	6 0 -1	-1 2 1	-2 -2 -2	-6 -3 -5	5 0 -1
812	Mississippi	X	к	-5 -1 7	-2 -4 0	1 0 -5	-1 -2 1	-6 -1 6	-2 -3 0	1 3 -6	-1 0 -1
824	Illinois	C	Х	5 6	-3 -2	-2 -3	-4 1	5 5	-2 -2	-3 -2	-3 -1
883	lowa	С	Х	5 7	-1 -1	2 -3	1 4	4 6	-1 -1	3 -1	2
1253	Ok1ahoma	х	A	-1 -9 0	-2 0 -2	-4 0 -5	-1 0 -1	0 -9 0	-2 0 -2	-2 0 -5	0 1 -2
1 502	Colorado	С	Y	-7 -8 1	-2 -3 -1	-5 1 -2	3 0 3	-7 -7 1	-2 -3 -1	-4 1 -2	2 1 3
1572	Nebraska	С	Р	-7 -5 -8	24 17 7	1 -2 0	-6 -7 2	-7 -5 -8	23 17 8	-3 -1	-5 -6 2

Symbol definitions:

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans

TABLE A-2.— VARIANCES C. THE PROPORTION ESTIMATES USING 60, 45, AND 30 STARTING DOTS

(a) Using 60 starting dots

		Clus	ter		Classify				
Segment number	Machine e	stimate	Stratifie estim		Machine 6	estimate	Stratified areal estimate		
	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	
1075	33.333	4.000	17.333	2.333	30.333	6.333	16.333	2.333	
1341	7.000	6.333	4.333	0.333	10.333	6.333	1.333	1.000	
1591	7.000	4.000	4.333	1.000	9.333	1.000	0.333	4.000	
185	2.333	4.000	1.000	5 . 3 33	2.333	4.000	1.000	4.333	
804	4.000	0.333	4.333	13.000	4.000	0.333	0.333	9.333	
812	1.000	1.333	28.000	1.000	1.000	1.333	32.333	0.333	
1253	0.333	0.0	12.333	0.333	0.333	0.0	9.333	1.000	
1502	14.333	5.333	9.333	4.333	14.333	5.333	9,000	1.000	
1572	19.000	14.333	0.333	20.333	21.333	10.333	0.0	30,333	
146	1.000	1.000	4.000	1.000	0.250	1.000	2.25 0	0,250	
824	2,250	1.000	9.000	12.250	2,250	1.000	4.000	9.000	
883	1.000	0.250	4.000	0.250	0.250	0.0	2.250	0.250	

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TABLE A-2.— Continued.

(b) Using 45 starting dots

		Clus	ter		Classify				
Segment number	Machine 6	estimate	Stratifie estim		Machine a	estimate	Stratified areal estimate		
·	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	
1075	24.333	2.333	8.333	4.000	23,333	2,333	14.333	4.000	
1341	26.333	9,333	4.333	1.333	26.333	6.333	3.000	0.333	
1591	12.333	12,333	0.333	4.000	16.333	9.000	0.333	2.333	
185	2.333	10.333	3.000	7.000	2.333	10.333	3.000	7.000	
804	0.333	1.000	1.000	9.333	0.333	1.000	1.000	10.333	
812	1.333	4.333	14.333	2.333	0.333	6.333	25.333	1.000	
1253	17.333	0.0	7.000	0.333	17.333	0.0	7.000	1.333	
1502	16,000	10.333	9.333	2.333	12.333	10.333	9.000	1.000	
1572	24.333	21.000	0.0	31.000	21.333	21.000	1.333	16.333	
146	6.250	6.250	0.0	2.250	4.000	6.250	1.000	1.000	
824	2.250	0.250	12.250	25.000	2.250	0.250	4.000	16.000	
883	2.250	0.250	2.250	0.250	2.250	1.000	4.000	0.250	

2 2

TABLE A-2.— Concluded.
(c) Using 30 starting dots

		Clus	ter		Classify				
Segment number	Machine e	stimate	Stratifie estin		Machine e	estimate	Stratified areal estimate		
	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	Crop 1	Crop 2	
1075	34.333	16.000	4.333	6.333	34.333	20.333	4.000	7.000	
1341	126.333	49.000	16.333	4.333	117.000	46.333	16.333	2.333	
1591	8.333	4.000	1.000	4.000	7.000	2.333	0.333	4.000	
185	6.333	1.333	6.333	3.000	6.333	1.333	6.333	5.333	
804	7.000	0.333	0.333	14.333	2.333	0.0	2.333	10.333	
812	36.333	4.000	9.333	1.333	37.333	2.333	26.333	0.333	
1253	24.333	1.333	7.000	0.333	27.000	1.333	6.333	2,333	
1502	24.333	1.000	9.000	3,000	21.333	1.000	6.333	1.000	
1572	2,333	73.000	2.333	24.333	2.333	57.000	4.000	19.000	
146	1.000	16.000	2.250	6.250	1.000	20.250	2.250	1.000	
824	9.000	2.250	2,250	6.250	12.250	2.250	0.0	2,250	
883	1.000	0.0	6.250	2.250	1.000	0.0	4.000	0.250	

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TABLE A-3.— VARIANCE OF THE PCC USING 60, 45, AND 30 STARTING DOTS

(a) Using 60 starting dots

Segment	Clus	ster	Clas	sify
number	PCC1	PCC2	PCC1	PCC2
1075	0.0	12.423	0.0	28.690
1341	0.963	0.333	0.963	4.680
1 591	0,963	5.230	0.963	3.253
185	0.0	5.590	0,0	3,243
804	0.0	4.103	0.0	4,823
812	0,963	2,170	0,963	10.663
1253	2.723	11.323	3,853	16.163
1502	0.963	2.803	0.963	0.010
1572	3,630	3.610	0.853	1.203
146	0.722	0.303	0.0	0.303
824	2.723	10.890	2.890	6,250
883	0.0	9.923	0.002	9.610

TABLE A-3.— Continued.

(b) Using 45 starting dots

Segment	Clus	ter	Clas	sify
number	PCC1	PCC2	PCC1	PCC2
1075	0.0	9.663	0.0	20.410
1341	1.613	2.613	1,613	2.823
1 591	11.590	0,093	41.070	0.120
185	6.750	9,390	1.763	3.773
804	0.0	2.203	1.613	2.203
812	6.453	8.423	5.063	2.203
1253	4.840	6.070	6.453	9.120
1502	0.0	5.590	0.0	6.413
1572	1.763	2.170	1.843	4.423
146	1.210	0.250	1.210	0.360
824	1.210	4.203	1.210	3.063
883	1.210	4.410	1.210	10.890

TABLE A-3.— Concluded.
(c) Using 30 starting dots

Segment	Clus	ter	Clas	sify
number	PCC1	PCC2	PCC1	PCC2
1075	0.0	13.890	0.0	34.443
1341	25.963	22.863	25.853	24.943
1591	3.630	2.403	3.630	5.643
185	3.743	19.253	27.543	23.543
804	0.0	5.813	0.0	3.803
812	14.963	7.680	3.853	2.110
1253	3,630	12.213	0.0	7.453
1502	0,0	11.543	0.0	13.043
1572	49.083	1.203	45.563	1.703
146	100.000	17.640	44.223	43,560
824	0.0	1.000	0.0	3.802
883	0.0	16.402	2.723	9,303

TABLE A-4.— THE R-VALUES FOR 60, 45, AND 30 STARTING DOTS USING LABELING PROPORTIONS

(a) R-values for 60 starting dots

	Segment	Crop 1	Crop 2	[Mean (µ)	Clust = 0,73852; v	er eriance (o ²) - 0.025]	[Mean (µ)	Class - 0.72888; v	ify eriance (o²) = 0.032]
Humber	Location	}	[[#] 10	*01	π	R	#10	" 01	π	R
1075	Hebraska	C	۸	0.033 .125 .067	0.600 .479 .444	0.429 .462 .429	0.786 .817 .707	0.067 .107 .050	0,622 .438 .409	0.429 .462 .423	0.852 .762 .645
1341	Kansas	c	х	.469 .408 .460	.023 .321 .345	.633 .533 .624	.666 .926 .961	.458 .429 .447	.250 .286 .345	.538 .533 .539	.910 .917 .955
1691	Nebraska	С	E	.059 .076 .111	.580 .577 .435	.167 .248 .221	.833 .835 .787	.071 .077 .062	.588 .560 .500	.168 .243 .214	.855 .824 .755
146	Kentucky	С	X	.154 .344	.155 .123	.690 .640	.561 .697	.154 .313	.172 .140	.690 .640	. 588 . 591
185	Minnesota	C	X	.043 .011 .034	.667 .867 .938	.114 .143 .152	.879 .934 .997	.044 .011 .023	.667 .867 .934	.118 .144 .154	.879 .934 .992
804	lowa	С	x	.269 .233 .194	.089 .107 .054	.752 .714 .705	.608 .573 .411	.269 .267 .194	.114 .096 .066	.752 .712 .072	.637 .586 .413
812	Mississippi	X	К	.191 .136 .250	.190 .197 .113	.652 .581 .505	.619 .666 .586	.174 .093 .250	.190 .180 .113	, 556 , 587 , 505	. 599 . 487 . 586
824	Illinois	C	X	.778 .500	.014 .011	.888 .898	.880 .617	.778 .500	.014 .011	.888 .897	.880 .616
883	Iowa	С	x	.308 .300	.243 .196	.729 .651	.828 .753	.269 .241	.267 .196	.729 .659	.813 .700
1253	Oklahoma	X	A	.016 .117 .183	.125 .178 .176	.377 ,429 .324	.419 .500 .622	.046 .102 .104	.103 .166 .147	.375 .433 .337	. 268 . 446 . 454
1502	Colorado	С	Y	.055 .060 .068	.714 .737 .412	.133 .181 .162	.921 .778 .716	.044 .000 .069	.667 .737 .412	.117 .186 .163	.878 .775 .717
1572	Nebraska	c	P	.636 .867 .455	.091 .013 .146	.733 .714 .790	.823 .935 .850	.571 .867 .455	.078 .013 .169	.733 .714 .790	.830 .935 .874

Symbol definitions:

R = reduction coefficient

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans

 $[\]pi$ = the probability a pixel is labeled wheat

 $[\]pi_{10}$ = the probability a pixel is classified wheat and labeled nonwheat

 $[\]pi_{01}$ = the probability a pixel is classified nonwheat and labeled wheat

TABLE A-4.— Continued.

(b) R-values for 45 starting dots

,	Segment	Crop 1	Crop 2	[Mean (µ)	Clust • 0,73476; v	er ariance (d ²) = 0.032]	[Mean (µ)	Class - 0.72915; v	ify ariance (o ²) - 0.030]
Number	Location			[#] 10	" 01	র	R	*10	[#] 01	Ħ	R
1075	Hebraska	С	A	0.050 .125 .100	0.578 .438 .378	0.429 .462 .429	0.795 .784 .695	0.068 .071 .067	0.591 .425 .386	0.427 .456 .423	0.831 .701 .651
1341	Kansas	С	x	.551 .367 .560	.232 .339 .345	.633 .633 .524	.947 .914 .991	.489 .347 .417	.218 .339 .309	.539 .533 .534	.907 .902 .924
1591	Nebraska	С	E	.106 .063 .099	. 588 . 654 . 522	.167 .248 .221	. 902 . 871 . 834	. 094 . 038 . 075	.647 .654 .500	.167 .248 .216	.921 .825 .780
146	Kentucky	C	x	.192 .219	.121 .140	. 690 . 640	.546 .594	.231 .250	.105 .105	.687 .640	.559 .571
185	Hinnesota	С	X	.065 .011 .079	.583 .800 .875	.114 .143 .152	.867 .880 .997	.067 .011 .057	.583 .800 .875	.119 .146 .155	.867 .880 .990
804	lowa	¢	X	.231 .200 .229	.127 .093 .054	.752 .714 .706	.624 .509 .452	.269 .233 .226	.128 .095 .055	.750 .712 .702	.661 .549 .450
812	Mississippi	X	K	.213 .182 .250	.172 .148 .132	. 552 . 581 . 505	.622 .554 .612	.196 .140 .235	.138 .230 .132	. 558 . 587 . 510	. 554 . 615 . 594
824	Illinois	С	X	.778 .400	.014 .011	.888 .898	.880 .519	.778 .300	.014 .023	.888 .896	.880 .496
883	Lowa	C	X	.192 .020	. 283 . 214	.625 .651	.741 .465	. 269 . 172	,232 ,196	.726 .659	.787 .629
1253	Oklahoma	х	A	.136 .067 .127	.100 .218 .176	.377 524 .324	.439 .483 .534	.109 .069 .132	.100 .244 .121	.385 .437 .327	.389 .501 .477
1502	Calorado	С	Y	, 077 , 000 , 080	.786 .684 .412	.133 .181 .162	.975 .725 .742	.079 .000 .080	.846 .684 .412	.127 .184 .163	.992 .726 .742
1572	Nebraska	С	Р	.714 .833 .545	.065 .013 .181	.733 .714 .790	.912 .910 .931	.679 .828 .545	.065 .014 .157	.733 .718 .790	.889 .908 .914

Symbol definitions:

R = reduction coefficient

 π = the probability a pixel is labeled wheat

 π_{10} = the probability a pixel is classified wheat and labeled nonwheat π_{01} = the probability a pixel is classified nonwheat and labeled wheat Crop codes

A - alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X - soybeans

TABLE A-4.- Concluded.

(c) R-values for 30 starting dots

Segment		Crop 1 Crop		[Mean (µ)	Clust • 0.77448;	er variance (o	²) = 0.029}	Classify [Keen (u) = 0.74633; variance $(o^2) = 0.034$]				
Number	Location	<u> </u>		" 10	[#] 01	Ħ	R	₹10	[#] 01	π	R	
1075	Nebraska	C	۸	0,050 .196 .163	0.689 .333 .356	0,429 .462 .429	0,877 .773 .779	0.050 .125 .133	0,682 .340 .302	0.423 .456 .417	0.872 .695 .667	
1341	Kansas	С	X	.510 .306 .680	.161 .339 .273	.467 .533 .524	.879 .875 .997	.511 .245 .688	.143 .321 .236	.544 .533 .534	.859 .812 .993	
1 691	Nebraska	С	E	.118 .076 .086	.706 .654 .739	.167 .248 .221	.966 .889 .952	.096 .064 .088	. 647 . 640 . 696	.170 .243 .223	. 922 . 863 . 931	
146	Kentucky	С	K	.385 .313	.165 .105	.690 .640	.783 .638	.462 .219	.158 .105	.687 .640	.845 .536	
185	Minnesota	C	x	.075 .044 .101	.833 .800 1.000	.114 .143 .152	.989 .952 .983	.056 .035 .112	.833 .800 .938	.119 .150 .162	. 900 . 938 . 997	
874	Iowa	C	х	.269 .300 .258	.177 .067 .041	.752 .714 .705	.734 .561 .452	.308 .276 .233	.143 .068 .027	.748 .718 .709	.719 .539 .386	
812	Mississippi	х	K	.170 .205 .308	.259 .115 .075	.552 .581 .505	.677 .631 .696	.109 .195 .294	.241 .197 .038	. 558 . 598 . 510	.582 .640 .519	
824	Illinois	C	x	.889 .500	.014 .045	.888 .898	.961 .772	.889 .400	.014 .047	.888 .895	. 961 . 694	
883	Iowa	C	x	.346 .267	.157 .086	.729 .659	.764 .558	. 280 . 241	.174 .070	.734 .663	.732 .497	
1253	Oklahoma	X	A	.121 .100 .127	.125 .289 .176	.377 .429 .324	.446 .604 .534	.108 .085 .132	.150 .289 .121	.381 .433 .327	.455 .580 .477	
1502	Colorado	С	٧	.066 .000 .045	.857 .684 .353	.133 .181 .162	.990 .725 .597	.069 .000 .048	.917 .684 .313	.121 .183 .162	.9996 .726 .569	
1572	Nebraska	С	р	.821 .867 .455	.000 .027 .205	.733 .714 .790	.862 .958 .904	.760 .862 .455	.000 .027 .181	.733 .718 .790	.804 .955 .885	

Symbol definitions:

R = reduction coefficient

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasture

X = soybeans

 $[\]pi$ = the probability a pixel is labeled wheat

 $[\]pi_{10}$ = the probability a pixel is classified wheat and labeled nonwheat

 $[\]pi_{01}^{\rm rot}$ = the probability a pixel is classified nonwheat and labeled wheat

TABLE A-5.— THE R-VALUES FOR 60, 45, AND 30 STARTING DOTS USING GROUND-TRUTH PROPORTIONS

(a) R-values for 60 starting dots

s	Segment		Crop 2	Cluster [Mean (μ) = 0.73779; variance (σ^2) = 0.024]				Classify [Mean (μ) = 0.73136; variance (o^2) = 0.034]				
Number	Location]		710	⁷⁷ 01	π	R	"10	^π o1	π	R	
1075	Hebraska	С	A	0,033 -125 -067	0.600 .479 .444	0.37	0.776 .815 .701	0.067 .107 .050	0.622 .438 .409	0.37	0.649 .758 .636	
1341	Kansas	c	х	.469 .408 .460	.023 .321 .345	,54	.664 .926 .962	.468 .429 .447	.250 .286 .345	.54	.910 .917 .956	
15}]	Nebraska	c	E	.059 .076 .111	.588 .577 .435	.22	.819 .840 .787	.071 .077 .062	.588 .560 .500	,22	.840 .829 .753	
146	Kentucky	С	×	.164 .344	.165 .123	.63	.540 .696	.154 .313	.172 .140	.63	.666 .690	
1 25	Minnesota	С	×	.043 .011 .034	.667 .867 .938	.16	.861 .932 .997	.044 .011 .023	.667 .867 .934	.16	.863 .932 .991	
804	lows	С	×	.269 .233 .194	.089 .107 .054	.75	.587 .588 .422	.269 .267 .194	.114 .095 .055	.75	.636 .598 .425	
812	Mississippi	x	к	.191 .136 .250	.190 .197 .113	.55	.619 .560 .581	.174 .093 .250	.190 .180 .113	.55	.599 .477 .581	
824	Illinois	С	×	.778 .500	.014 .011	.93	.898 .640	.778 .500	.014 .011	.93	.997 .640	
883	lows	C	х	.308 .309	.243 .196	.70	.820 .765	.269 .241	.257 .196	.70	.803 .713	
1253	Ok 1 ahoma	x	A	.106 .117 .183	.125 .178 .176	.36	.423 .509 .610	.046 .102 .104	.103 .156 .147	.36	.269 .465 .448	
1502	Colorado	c	Y	.055 .030 .068	.714 .737 .412	.15	.917 .767 .723	.044 .000 .069	.667 .737 .412	.15	.866 .767 .726	
1572	Nebraska	С	P	.536 .867 .465	.091 .013 .145	. 7/3	.828 .934 .840	.571 .867 .455	.078 .013 .169	.78	.839 .935 .871	

Symbol definitions:

R = reduction coefficient

Crop codes

A = alfalfa

C = corn

E = sorghum

K = cotton

P = pasturo

X = soybeans

 $[\]tau$ = the probability a pixel is labeled wheat

 $[\]pi_{01}$ = the probability a pixel is classified wheat and labeled nonwheat

 $[\]pi_{10}^{-}$ = the probability a pixel is classified nonwheat and labeled wheat

TABLE A-5.— Continued.

(b) R-values for 45 starting dots

	(agment		Crop 2	Cluster [Hean (μ) = 0.73579; variance (σ^2) = 0.032]				Classify [Mean (μ) = 0.72982; variance (σ^2) = 0.029]				
Number	Location			"10	^π O1	π	R	"to	[™] 01	π	R	
1075	Nebraska	С	A	0.050 .125 .100	0,578 .438 .378	0.37	0, 788 . 782 . 693	0,068 .071 .067	0.591 .426 .386	0,37	0,827 ,691 ,646	
1341	Kansas	С	×	.551 .367 .560	.232 .339 .345	.54	.947 .914 .991	.489 .347 .417	.218 .339 .309	.54	.907 .902 .924	
1591	Nebreska	С	E	.106 .063 .099	.588 .654 .522	.22	.888 .875 .835	.094 .038 .075	.647 .654 .500	.22	.910 .828 .779	
146	Kentucky	С	×	.192 .219	.121 ,140	.63	.531 .592	.231 .250	.105 .105	.63	.549 .570	
185	Hinnesota	С	Х	.065 .011 .079	.583 .800 .875	.16	.844 .879 .996	.067 .011 .057	.583 .800 .875	.16	.847 .879 .990	
804	Lowa	C	x	.231 .200 .229	.127 .093 .054	.75	.623 .524 .461	.269 .233 .226	.128 .095 .055	.75	.661 .563 .460	
812	Hississippi	х	к	.213 .182 .250	.172 .148 .132	.55	.622 .551 .609	.196 .140 .235	.138 .230 .132	.55	.554 .607 .592	
824	Illinois	c	×	.778 .400	.014 .011	.93	.898 .544	.778 .300	.014 .023	.93	.898 .544	
883	Iowa	С	x	.192 .020	.283 .214	.70	.765 ,501	.269 .172	.232 .196	.70	.778 .646	
1253	Ok lahowa	Х	۸	.136 .067 .127	.100 .218 .176	.36	.444 .463 .524	.109 .069 .132	.100 .244 .121	.36	.396 .497 .465	
1502	Colorado	С	٧	.077 .000 .080	.786 .684 .412	.15	.973 .718 .750	.079 .000 .080	.846 .684 .412	.15	.991 .718 .750	
1572	Nebraska	С	P	.714 .833 .545	.065 ,013 ,181	.78	.91 <i>7</i> .909 .930	.679 .828 .545	.065 .014 .157	.78	.895 .908 .912	

Symbol definitions:

R = reduction coefficient

Crop codes

A - alfalfa

€ • corn

E * sorghum

K = catton

P = pasture

X = soybeans

 $[\]pi$ = the probability a pixel is labeled wheat

 $[\]pi_{10}$ - the probability a pixel is classified wheat and labeled nonwheat

 $[\]pi_{O1}$ = the probability a pixel is classified nonwheat and labeled wheat

TABLE A-5.— Concluded.

(c) R-values for 30 starting dots

٠	Segment		Crop 2	(Mean (µ) =	Clusto 0,77509; v	er eriance (o	²) = 0.029]	Classify [Mean (μ) = 0.74636; variance (u^2) = 0.034]				
Number	Location			¹¹ 10	[#] 01	T	R	"10	TO1	π	R	
1075	Nebraska	С	٨	0,050 ,196 ,183	0,689 ,333 ,356	0.37	0,873 .778 .763	0.050 .125 .133	0.682 .340 .302	0.37	0,868 .695 .869	
1341	Kansas	С	х	.510 .306 .680	.161 .339 .273	.54	.875 .875 .997	.511 .245 .6/8	.143 .321 .236	,54	.859 .813 .993	
1591	Nebraska	C	E	.118 .076 .086	.705 .654 .739	.22	.960 .893 .952	.096 .064 .088	.647 .640 .696	.22	.912 .866 .932	
146	Kentucky	С	X	.385 313	.155 .105	.63	.775 .637	.462 .219	.158 .105	.63	.839 .535	
185	Minnesota	С	x	.075 .044 .101	.833 .800 1.000	.16	.986 .949 .902	.056 .035 .112	.833 .800 .938	.16	.976 .937 .996	
804	LOWA	С	X	.269 .300 .258	.177 .067 .041	.75	.734 .569 .456	.308 .276 .233	.143 .068 .027	.75	.720 .547 .385	
812	Hississippi	X	K	.170 .205 .308	.259 .115 .075	.55	.677 .531 .587	,109 .195 .294	.241 .197 .038	.55	.580 .633 ,507	
824	Illinois	Ç	X	.889 .500	.014 .045	.93	.970 .810	.889 .400	.014 .047	.93	.970 .746	
883	lowa	¢	X	.346 .267	.157 .086	.70	.756 .564	.280 .241	.174 .070	.70	.720 ,502	
1253	0k1ahoma	X	۸	.121 .100 .127	.125 .289 .176	.36	.451 .605 .524	.108 .085 .132	.150 .289 .121	,36	.459 .578 .465	
1502	Colorado	С	Y	.066 .000 .045	.857 .604 .353	.15	.989 .718 .605	.069 .000 .048	.917 ,864 .313	.15	.9996 .718 .577	
1572	Hebraska	C	P	.821 .867 .455	.000 .027 .205	,78	.855 .960 .902	.750 .862 .455	.000 .027 .181	.78	.794 .957 .882	

Symbol definitions:

R = reduction coefficient

r = the probability a pixel is labeled wheat

 \mathbf{m}_{10} = the probability a pixel is classified wheat and labeled nonwheat

*01 - the probability a pixel is classified nonwheat and labeled wheat

Crop codes

A - alfalfa

C = com

E = sorghum

K = cotton

P = pasture

X = soybeans