Utilizing Remote Sensing of Thematic Mapper Data to Improve Our Understanding of **Estuarine Processes and Their** Influence on the Productivity of **Estuarine-Dependent Fisheries**

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Final Report to the National Aeronautics and Space Administration

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Final Report to the National Aeronautics and Space Administration Goddard Space Flight Center Greenbelt, Maryland 20771

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Abstract

The land-water interface of coastal marshes may influence the production of estuarinedependent fisheries more than the area of these marshes. To test this hypothesis, we created a spatial model to explore the dynamic relationship between marshland-water interface and level of disintegration in the decaying coastal marshes of Louisiana's Barataria, Terrebonne, and Timbalier basins. Calibrating our model with Landsat Thematic Mapper satellite imagery, we found a parabolic relationship between land-water interface and marsh disintegration. Aggregated simulation data suggest that interface in the study area will soon reach its maximum and then decline. We found a statistically significant positive linear relationship between brown shrimp catch and total interface length over the past 28 yr. This relationship suggests that shrimp yields will decline when interface declines, possibly beginning about 1995.

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Introduction

The loss of Louisiana's coastal wetlands at the average rate of about 100 km²/yr (Gagliano et al. 1981) is a problem of national concern because of their importance to wildlife and fisheries. Louisiana's seafood production, the highest in the nation, is based on species dependent on estuaries and their associated wetlands, which provide food and shelter for young organisms (Boesch and Turner 1984).

Both natural processes and human interference with these processes are responsible for the rapid wetland loss in Louisiana (Baumann et al. 1984). The leveeing of the Mississippi River has prevented the deposition of marsh-building sediment that could offset subsidence and sea-level rise (Kesel 1988). Drainage and navigation channels have altered the natural hydrologic processes that build coastal and interior marshes and stimulate marsh vegetation growth (Turner and Cahoon 1987).

Despite the loss of wetlands and the known dependence of fishery species on wetlands, statistics indicate that Louisiana's fishery landings have been increasing. The increase in landings, not fully explained by an increase in effort (Nichols 1984), has created a sense of false security that has delayed action to curb wetland loss.

The production of fishery species may be more dependent on the land-water interface than on wetland acreage. Faller (1979), Dow (1982), and Gosselink (1984) found statistically significant relationships between fishery production and land-water interface in neighboring areas. Zimmerman et al. (1984) noted that brown shrimp densities were highest in areas of high shoreline "reticulation."

Using a stochastic computer model, Browder et al. (1984) provided a theoretical description of how the length of the land-water interface changes during marsh disintegration. They found that interface length increased in early stages of simulated marsh disintegration, reached a maximum when the marsh was roughly 50% water, and decreased thereafter. They further noted that the magnitude of maximum interface was variable and was affected by the spatial pattern of land and water--specifically the degree of clumping of water pixels to form water bodies.

In the study reported here, we refined and expanded the Browder et al. (1984) model and calibrated it with Thematic Mapper (TM) imagery covering 70 marsh sites in coastal Louisiana (Appendix A). Then we used our model to simulate the complete cycle of marsh disintegration

at each site and collected data on interface length. We used independent data to roughly convert interface length versus disintegration level to interface length loss versus time at each site. Then we tested total interface length from the 70 simulations for its ability to explain annual brown shrimp catch in estuaries adjacent to the study area. Finding a statistically significant relationship, we used it to estimate future shrimp production. We compared data from the TM imagery to simulated data from our model in order to evaluate Browder et al.'s general observations concerning the relationship of interface length to land loss and the spatial pattern of land and water.

Methods

The study can be viewed as consisting of four steps: model development, model calibration, model evaluation, and model extension. Model development consisted of refining the Browder et al. (1984) model for use with TM data. The model contains three adjustable parameters that were calibrated to the spatial patterns of land and water in 70 marsh sites in Louisiana, as indicated in TM imagery. The TM scenes were classified into land and water pixels, and several measures, or indices, of spatial pattern were obtained from each scene. We developed an expert system that used these spatial-pattern indices to select the model-parameter values to simulate the history of marsh disintegration at each site. Then we made 70 "best fit" simulations of marsh disintegration-one for each scene--and recorded the history of interface length as a function of disintegration level (DL, water area as percentage total area) throughout each simulation.

We evaluated our simulations by several methods. We used regression analysis to compare the spatial-pattern indices of the 70 simulations to those of the TM scenes. We compared the number of water-body groups in the simulation to those in the TM scene, by lobe and marsh type. We visually compared the TM scenes with our simulated scenes at the same stages of disintegration. Finally, we examined the model-parameter values selected by the expert system, comparing them by marsh type and lobe.

We extended our model results to fishery production. First we determined the relationship between brown shrimp catch and simulated interface length in the study area for the past 28 yr. Finding a statistically significant relationship, we used it to estimate future shrimp catches based on simulated future interface length.

Description of the New Model

Our new model simulates marsh disintegration by successively changing land pixels to water pixels. The relative probability that a land pixel will be converted to water at each iteration is governed by a function weighted by three adjustable parameters: interior disintegration (W), shoreline erosion (G), and border-condition (BC). The weighting parameters were based on Sasser et al.'s (1986) observation that two patterns of marsh disintegration occur in Louisiana. In one pattern, small, randomly spaced, gradually expanding water bodies develop in solid marshes. In the other, land disappears along the margins of major water bodies, as if lost to waves or other erosive forces. The model simulates the entire disintegration process, starting with solid land and ending with only open water. Each iteration represents passage of time, although time units are unspecified.

The pixel to be disintegrated at each iteration is selected from a numbered list by a pair of randomly generated numbers (RN). The first makes a tentative selection by matching a number on the list; the second random number determines whether the tentative selection is eligible. The pixel is eligible if its total weight at that iteration $(F_{i,j,k})$ is greater than $RN_k \cdot max-F_{i,j}$ for that k (i and j = pixel coordinates and k = iteration). The relative probability that a specific pixel will be selected at iteration k $(RP_{i,j,k})$ is the ratio of the total weight of that pixel to the sum of the total weights of all the land pixels:

$$RP_{i,j,k} = F_{i,j,k} / \sum_{i=1}^{r} \sum_{j=1}^{c} (F_{i,j,k}),$$
(1)

where r = number of rows, c = number of columns,

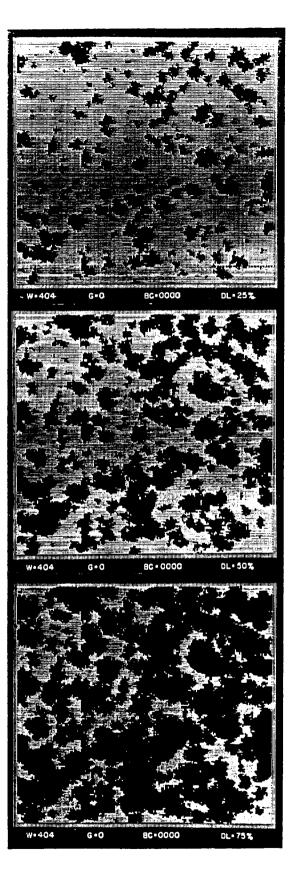
$$F_{i,j,k} = 1 + W \cdot S_{i,j,k} + \sum_{m = 1} [BC_m \cdot G_m \cdot B_{m,i,j,k}].$$
(2)

W = weighting coefficient for each side of the pixel adjacent to water. $S_{i,j,k}$ = number of pixel sides adjacent to water. G = weighting coefficient for pixels bordering a major water body, by border. B = a Boolean value (1 or 0) indicating whether the pixel is on a major water body, by marsh border. Border condition is the vector BC. BC_m indicates which marsh borders (n, s, e, or w) are on major water bodies. (Note that throughout this paper, "side" refers to *pixel*

boundary and "border" refers to *marsh* boundary.) Once a pixel is converted to water, it is removed from the selection list, shortening it by one. Figure 1 gives a snapshot view of the progress of marsh disintegration in one simulation. The marsh is initially solid. By the time it is 25% disintegrated (DL = 25%), we see many small water bodies. Water areas are larger and are beginning to coalesce at 50% disintegration. Most water bodies are connected by the time the marsh is 75% disintegrated.

At each iteration of the model, counters keep track of the percentage of the total area that is water and the length of the land-water interface. Percentage water area is referred to throughout this discussion as the level of disintegration (DL). Land-water interface is measured in pixel-lengths--the length of one side of the square pixel. As measured, interface is homologous to the "join" statistic of Moran (1948) and is related to other spatial autocorrelation statistics indicating degree of clumping of the same pixel types (Upton and Fingleton 1985). By affecting the order of pixel disintegration, our model's weighting coefficients determine the degree of clumping of water pixels in simulated marshes. Figure 2 shows two marshes at similar stages of decay simulated by different interior-marsh-decay weighting coefficients. Note that water bodies are larger when W = 3,184 (bottom) than when W = 248 (top). (The erosion weighting coefficient was zero for both.)

The new model differs from the Browder et al. (1984) model in several important details. In the original model, only pixels *initially* on a major water body had the G-weighting (B = 1). The G-effect was inconsequential in sensitivity tests, particularly as the size of the simulated marsh increased. In the new model, any pixel can eventually be assigned B = 1, if it is connected to a designated water border by a continuous water path. The G-parameter now has a much greater effect. The new model allows flexibility in the initial identification of water borders, and up to four water borders can be set. The pixel-selection procedure of the new model is an improvement that made it practical to simulate marshes having as many pixels as the TM images of our study sites, 192 x 192. Appendix B presents the spatial-pattern statistics of simulations using all combinations of W, G, and BC. The new model and all ancillary programs were written in C and executed on an AT&T PC-7300, a 16-bit, 10-MHz computer with a Unix-V operating system.



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Figure 1. Snapshot view of simulated marsh disintegration at the 25%, 50%, and 75% disintegration levels (percentage open water area).

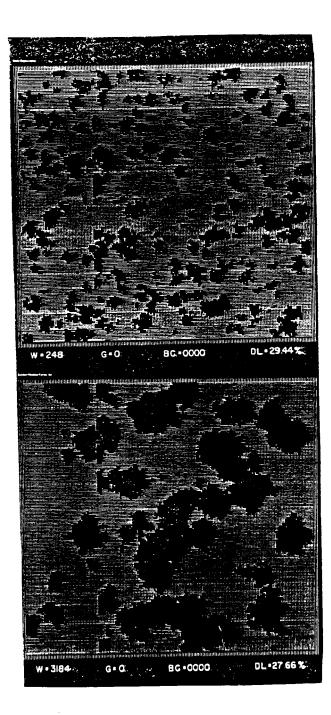


Figure 2. Two simulated marshes at similar levels of disintegration, produced by different values of W, the interior-marsh-decay weighting factor.

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Model Calibration

TM image processing. We analyzed the TM scenes on the Fisheries Image Processing System (FIPS) of the National Marine Fisheries Service (NMFS) in Slidell, Louisiana, and a system operated by the Florida Department of Natural Resources in St. Petersburg. Both systems consisted of a minicomputer, color-image display device, and other hardware for processing remotely sensed digital data. The software was a modified version of the Earth Resources Laboratory Applications Software (ELAS) (Graham et al. 1984).

The TM image we used was from a 2 December 1984 Landsat-5 overflight (Scene ID: 50276016022). Covering most of the Mississippi deltaic plain, it was one of the few relatively cloud-free images of our study area (quads 1 and 2 in path 22 and row 40 of the World-Wide Reference System).

ELAS modules PMGC (Georef constants-EROS format) and PMGE (Georef-EROS format) (Graham et al. 1984) were used to digitally rotate the images to fit a Universal Transverse Mercator projection with a north-south orientation. We used these modules to accumulate ground-control points, generate polynomial least-squares mapping equations, and resample the image with bilinear interpolation. Registration accuracies averaged 22-56 m. Resolution was the length of a TM pixel side, 30 m. Land and water pixels were classified by multiplying bands 4 and 5 (0.76-0.90 μ m and 1.55-1.75 μ m, respectively), rescaling to 0-255, and applying Pun's (1981) global thresholding technique.

Study-site selection. The study sites are located on two abandoned delta lobes of different ages. The early Lafourche lobe was an actively prograding delta within the last 1,800 yr. The late Lafourche lobe was an active distributary of the river within the last 600 yr. Chabreck (1972) distinguished four major types of Louisiana coastal marsh on the basis of vegetation: salt, brackish, intermediate, and fresh. Salt and brackish marshes are the most important marshes to estuarine-dependent fishery species and show a wide range of decay stages. For these reasons, we limited our study to these two more seaward marsh types.

Site locations are within the areas represented by 21 U.S. Geological Survey 7.5-min topographic maps. We used these maps and a coastal habitat map (Chabreck and Linscombe 1978), coupled with our extensive field experience, to distinguish brackish and salt marsh. We defined potential boundaries of study sites by dividing the area of the TM image corresponding

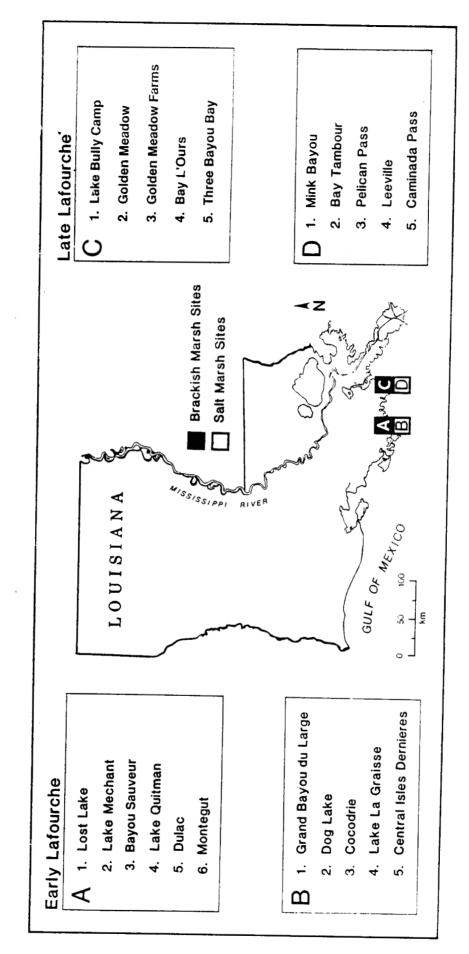
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to each topographic map into four contiguous quarters measuring 192 x 192 pixels (5,760 x 5,760 m, roughly 33.18 km²). The intersection of the quarters corresponded to the center of the map. We selected 70 marsh sites: 38 salt (20 on one lobe, 18 on the other) and 32 brackish (19 and 13 per lobe) (Fig. 3).

Measurement of spatial-pattern indices. We generated 70 binary land-water images from the band-4-x-band-5 images. To measure our spatial-pattern indices, we tabulated the following using ELAS command strings: (1) number of land and water pixels (to determine percentage water area = disintegration level [DL]); (2) number of water pixels by scan line and element column (to determine border condition); (3) number of land-water pixel-side contacts (interface length); (4) number of water pixels, excluding border pixels, with sides adjacent to zero, one, two, three, or four other water pixels (which we will refer to hereafter as the "side-adjacency" statistics); and (5) number of pixels in each water body (water-body size). Diagonal, or corner, contacts by water pixels were considered to connect two parts of the same water body.

We tabulated interface length in a three-step process. First, we generated an intermediate image using the ELAS shoreline-length (SLIN) module (Graham et al. 1984). SLIN uses a 3-x-3-pixel moving-window technique to classify each *land* pixel adjacent to water into 1 of 69 shoreline categories (Dow, 1982; Dow and Pearson, 1982). Second, we used a look-up table to convert the SLIN image to an image file of six classes: land and water pixels and *land* shoreline pixels having one, two, three, or four sides adjacent to water. Our principal spatial-pattern index, interface, was determined by counting the *land*-pixel sides adjacent to water pixels. We determined the number of *water*-pixel sides adjacent to other water pixels with a similar technique to obtain the side-adjacency statistics, which were our other major indices of spatial pattern. Two processing changes were required: *water* pixels adjacent to land were defined as *water* shoreline pixels in SLIN-module processing; and a new look-up table was used to classify *water* pixels with zero, one, two, three, or four sides adjacent to other water pixels. The water-body classifier (WBOD) of ELAS was used to determine water-body size.

The length of an irregular shoreline is a function of measurement unit (Mandelbrot 1967). Our measurements of land-water interface and, possibly, other spatial-pattern indices are valid



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Figure 3. Map of Louisiana showing study area and corresponding U.S. Geological Survey topographic maps covering the area.

only at TM resolution, 30 m. Appendix C presents the spatial-pattern statistics of the TM scenes.

Calibrating the model. An expert system, consisting of a knowledge base and decision process, selected the model parameters W, G, and BC to best approximate the spatial patterns of each study site. Selections were made by matching certain spatial-pattern indices of the imagery to those in a knowledge base built from simulations. Interface length, the five side-adjacency statistics, and a "target" border condition were the indices to be matched. The knowledge base showed how these variables changed in model executions as functions of W, G, and BC. The decision process consisted of rules for selecting the best W-G-BC combination.

The knowledge base was built by running simulations with all possible W and G combinations from the set [0, 4, 20, 60, 180, and 540] for the six types of BC. (Throughout this report, BC is given six possible Boolean values: 0000, 0001, 0011, 0101, 0111, and 1111, which show the specific spatial relationships of the borders, 0 indicating land and 1, water.) For BC = 0000, the set was extended to include W = 1,620 and 9,720. (In addition, power functions extrapolated to larger W's.) Target BC is determined by comparing DL to the percentage water pixels (P_m) in each row or column forming the outer border of the marsh, as follows:

Target
$$BC_m = 1$$
 if $P_m > DL$. (2)

The spatial-pattern indices of simulated marshes at the same DL as the study site were used to obtain one or more weighted mean W for every G-BC combination. For each G-BC, there could be one or more W based on each spatial-pattern index. To calculate the mean, W's were weighted by the number of water pixels of the index. For instance, if Adj-4 = 1,940, the weight given to the W obtained by matching this index was 1,940. The weight given to the W obtained by matching interface length was the sum of all water pixels. The weights used to calculate mean W were summed to calculate a "decision number" (DE) for each weighted mean W. DE was used to select the best W-G-BC from the many alternatives calculated for each site.

Another criterion used to select the best W-G-BC was coefficient of variation of the weighted mean W (CV). CV was a useful criterion because low CV indicated a high degree of convergence of W's estimated from all contributing spatial-pattern indices.

BC was the main criterion used to select the best W-G-BC combination. If the target BC was not matched by a solution meeting other criteria, the solution having BC most similar to the

target was selected. In our 70 cases, BC usually matched target BC or differed by only one border. The decision algorithm selected the W-G-BC combination having, first, BC most similar to target BC; second, high DE (within at least 75% of the highest DE among all alternative W-G-BC combinations); and third, lowest CV (see Appendix D, F).

Simulation of study-site spatial patterns. Once selected, model parameters were used to simulate the spatial pattern of each study site and the change in land-water interface with land loss. The land-water maps and spatial-pattern indices of the 70 simulated marshes were captured at the same levels of disintegration as corresponding study sites. In addition, interface length was recorded at each 5% level of disintegration as the simulation proceeded from solid land to open water.

Analysis of brown shrimp catch data. To relate marsh-water interface to annual fishery catch data, we needed to estimate interface length as a function of time. Interface length in our model output was expressed as a function of DL, not time. Therefore, we needed an estimate of the time trend in DL. We used data from Wicker's 1956 and 1978 maps (1980) to estimate this trend. The data were compiled by Liebowitz (Louisiana State University, private communication, 1988), who provided us with water area for each topographic-map area corresponding to our study sites. We estimated average annual change in DL per topographic-map area by expressing water area in 1956 and 1978 as percentage total area and calculating the annual average of the difference. This assumed a linear trend in water area from 1956 to 1978, which we projected into the future. We aggregated the data for each site to obtain, for each lobe, an estimate of interface, by year, from 1956 until the future total loss of marsh and interface. (An in-depth comparison of 1956 and 1978 data from the Wicker maps is presented in Liebowitz and Hill [1988].)

Using regression analysis, we compared 1960-1987 of the simulated interface time-series with unpublished brown shrimp catch data for Barataria, Timbalier, and Terrebonne bays for the same period (G. Davenport, NMFS, Miami, personal communication, 1988) to estimate a relationship between catch and interface. (Barataria Bay is associated with the late Lafourche lobe, and Timbalier and Terrebonne bays are associated with the early Lafourche lobe.) We predicted future shrimp catches from this relationship. Included as independent variables in the analysis were local rainfall (R. Muller, Louisiana State University, personal communication, 1988) and number of hours from April 9 through 30 in which temperatures were below 20°C (Barrett and Gillespie 1975; B. Barrett, Louisiana Department of Wildlife and Fisheries, personal communication, 1988). Barrett and Gillespie (1975) suggested that salinity and the temperature variable affected brown shrimp catches. We used rainfall as an inverse surrogate for salinity. Lack of reliable effort data precluded inclusion of this variable in our analysis.

Results

Results are organized as (1) interface length versus disintegration level of study sites, (2) evaluation of simulations, (3) simulated site-specific interface length versus DL, (4) aggregated interface length versus year, and (5) possible impacts on fisheries.

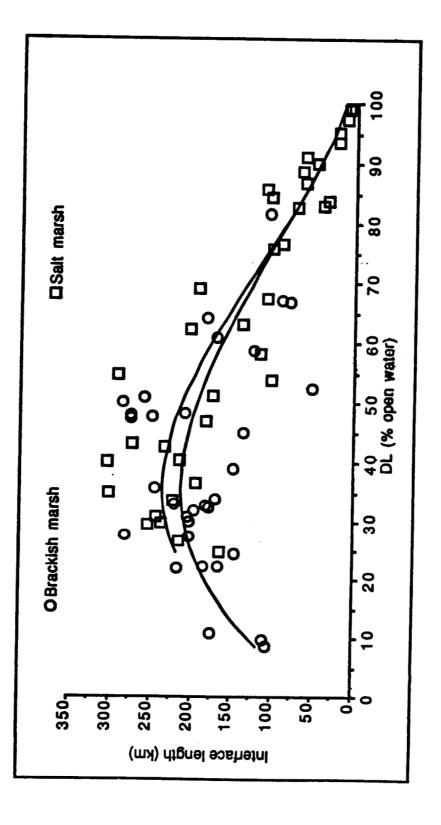
Interface Length versus Disintegration Level of Study Sites

A plot of interface length versus DL measured in the classified imagery of the study sites (Fig. 4) suggests that interface increased in the early stages of disintegration, reached a maximum when marshes were 30%-50% disintegrated, and decreased thereafter. Statistically significant (p < .1) parabolas were fit to separate data for salt and brackish sites. Most salt marsh sites were more than 50% disintegrated, whereas the DL of brackish sites ranged from low to high. DL and interface length did not differ significantly between early and late lobes, possibly because we excluded open-water areas of both lobes from our analysis.

Evaluation of Simulations

Following are the results of our evaluations of how well the simulations represented the spatial patterns of the study sites. Appendices E and G provide further specific comparisons in tabluar and graphic formats.

Agreement of simulation and study-site interface. Interface length in each simulation was obtained at the same DL as the TM scene it represented. Then the 70 simulation interfaces were regressed on the corresponding TM-scene interfaces. TM-scene interface explained 94% of the variation in simulation interface. The slope of the relationship was 1.06. The greatest



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Figure 4. Interface length versus disintegration level (percentage open water area).

departures of simulation from TM-scene interfaces were in the highest values. In most departures, the simulation value was higher than the TM-scene value. Half the simulation interface lengths differed from corresponding TM-scene values by no more than 10%, and 86% differed by no more than 30%. The average absolute difference was 11.7%.

Regression of simulation side-adjacency statistics on their TM-scene corollaries indicated highly significant relationships (F-stat. p < .001) for all but Adj-0, with 56%-99% of the variation in the simulation values explained by TM-scene values. R²'s were 0.56 for Adj-1, 0.91 for Adj-2, 0.80 for Adj-3, and 0.99 for Adj-4. Their slope coefficients varied from 0.97 to 1.21. The poor fit of simulation Adj-0 to TM-scene Adj-0 probably was largely due to the usually low value and resultant extremely small influence of this spatial-pattern index in the decision process.

Variation in simulation indices. Three replicate simulations with three sets of model-parameter values revealed the variation in simulation spatial-pattern indices caused by the random aspect of the model. CV averaged across all the spatial-pattern indices ranged from 4.9% to 19.3%. It was highest in the three replicate simulations where G = 540 and lowest in those where G = 0. The CV of Adj-0 was extremely high in the sets of replicate simulations in which G = 180 (CV = 53%) and G = 540)CV = 71%), probably because of the low value of Adj-0 (less than 20 pixels in all cases). Average CV's for the other side-adjacency statistics ranged from 2.17% for Adj-4 to 7.34% for Adj-2. The CV of interface length averaged 6.28%.

Water-body size groups. Water-body size data for sites and simulations were difficult to compare because water bodies were few and their size range enormous. Rather than grouping them by even intervals, we defined breaks between size groups with the following consistently applied algorithm. In a list of water bodies sorted by size, a break was defined if the larger of two adjacent water bodies was more than twice the size of the smaller. Upper and lower boundaries were placed on water-body size groups for each marsh unit. Figure 5 summarizes the differences between number of water-body size groups in each case. The study sites and their simulations had the same number of water-body groups in 23 of 70 cases. In 59 cases, study sites and simulations differed by no more than one group. This was good agreement considering that the status of one pixel in a strategic location could determine whether two

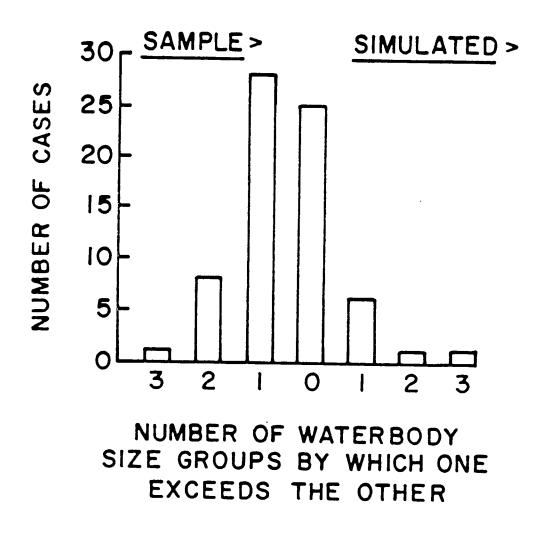


Figure 5. Frequency distribution of cases by difference between number of water-body size groups. Values on the abscissa indicate the number of groups by which the study site exceeds the simulation (values to the left of zero).

clumps of pixels formed one water body or two. Usually, the study site had more groups than the simulated marsh. The average number per study site was 2.7, whereas the average number per simulated marsh was 2.3. Two groups were distinguished for most marshes. Typically, a small percentage of water pixels were distributed among many small water bodies, and the rest were in one large water body. For example, in one study site, 4.3% of the water pixels were in water bodies that included 0.003%-0.752% of the total water pixels, whereas 95.5% were in one water body. Water-body groups in the corresponding simulation were similar. Generally, when more than two water-body groups occurred, the additional ones were at the lower end of the size range.

Visual evaluation. Visual comparisons suggested that the model often succeeded in simulating spatial patterns of the TM scenes, except when high G-values were used to simulate brackish marshes. The simulations did not appear to accurately represent those patterns of land and water heavily influenced by underlying geologic features, such as ridge/swale topography or large lakes, nor man-made features such as canals and diked areas. Despite limitations, the model simulated the general patterns of most marshes well, and matched a few remarkably well. The marsh map in Figure 6 (bottom) was simulated with an interior-marsh-decay coefficient of 311, a shoreline-erosion coefficient of 540, and a BC of 0001. At a DL of 68.89%, it displays a spatial pattern of land and water very similar to that of the classified TM scene at the same decay stage (Fig. 6, top). Interface length in the simulation map differed from that in the TM scene by 10.2%.

Model-parameter values. Some generalizations can be made about the appropriate modelparameter values for simulating marsh disintegration. Lobe age did not appear to influence parameter values, whereas marsh type seemed to be an influencing factor. Based on the knowledge base and our criteria, the expert system gave salt marshes higher shoreline-erosion coefficients, more water borders, and lower interior-decay coefficients than it gave to brackish marshes (Fig. 7). W and G were inversely related in the expert system's selections (Fig. 8A). Our visual comparisons suggested that low-to-medium values of G (0-180) and moderately high values of W (about 200-400) matched the spatial patterns of brackish TM scenes best. Conversely, high G-values (180 and 540) gave the best match to salt marsh scenes. Because salt

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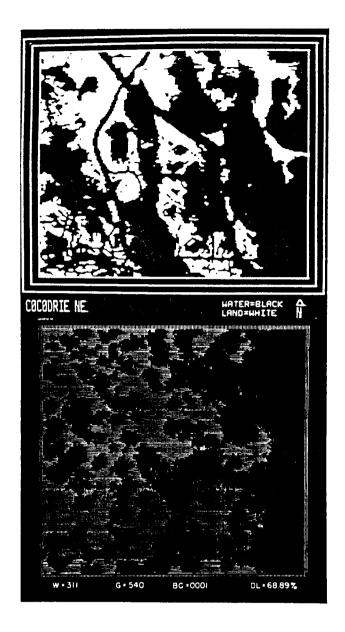


Figure 6. Land-water image of a study site (top) and its simulation (bottom) showing disintegration level (DL) and model coefficients (W, G, and BC) used in the simulation. (In BC, 0 = land and 1 = water.)

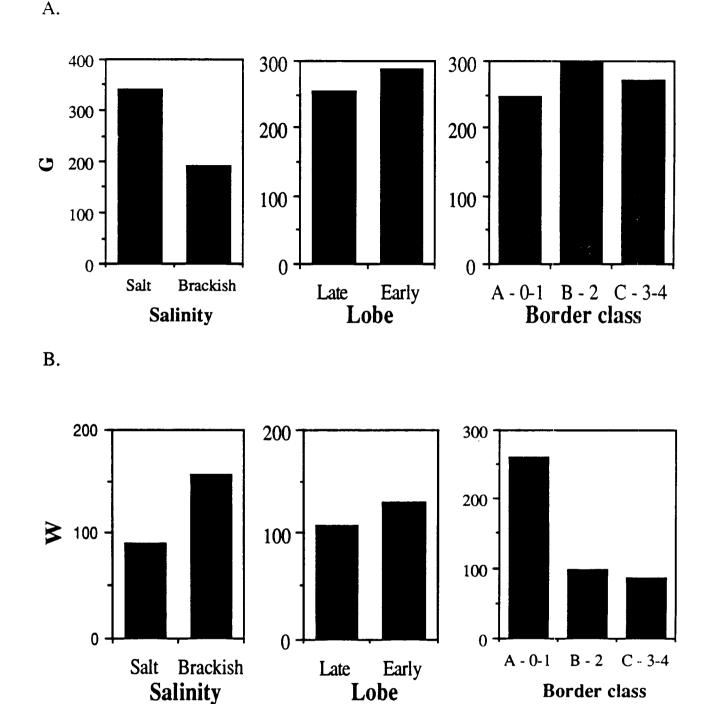
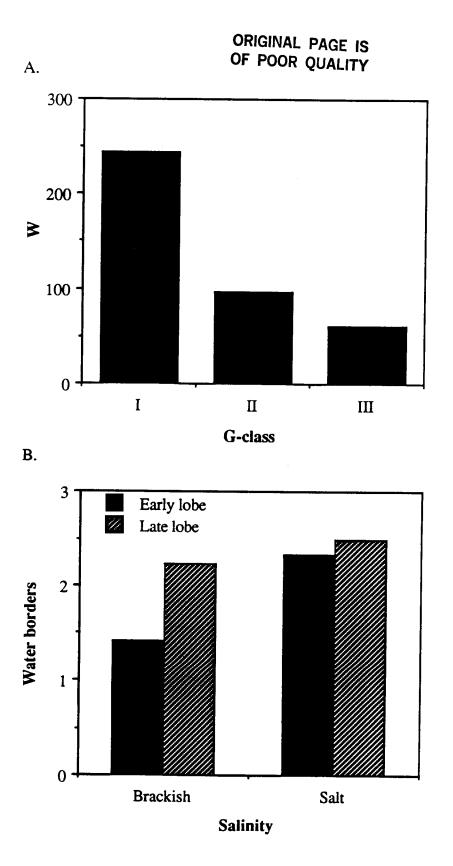


Figure 7. (A) Average G weighting factor (shoreline erosion) and (B) average W weighting factor (interior-marsh-decay) selected by the expert system, by marsh type, lobe (early or late Lafourche), and border condition (BC). (Two extremely high outliers were excluded in cacluclating average W.)





(A) Average W by G-class (Class 1, G = 0, 4, 20, or 60; Class 2, G = 180; Class 3, G = 540).

(B) Average border condition (BC) selected by the expert system, by marsh type and lobe.

marshes have more borders on major water bodies, shoreline erosion is more prevalent in them than is interior decay (Fig. 8B). Interior decay weighting coefficients selected by the expert system were highest for simulated marshes having the fewest water borders (Fig. 7B), which were primarily the brackish marshes (Fig. 8B).

Simulated Site-Specific Interface Length versus DL

We followed interface from 100% land to 100% water in each of the 70 simulations. Many simulations were similar to those of Lost Lake NW, Mink Bayou SW, and Mink Bayou SE (Fig. 9). Interface reached a maximum approaching 10,000 pixel-lengths (300 km) when the marsh was roughly 50% disintegrated. Interface in Pelican Pass SW (Fig. 9) followed a strikingly different path, reaching its unusually low maximum of 2,417 at a DL of only 11%. This is one of two simulations that differed markedly from the rest in reaching maximum interface at a low DL. Both were simulated with G = 540 and BC = 1111. The distribution of maximum interface in the 70 simulations was bimodal, with a lower peak around 2,000-4,000 pixel-lengths and a higher one at 9,000 (Fig. 10A). DL at maximum interface was between 45% and 60% in most simulated marshes (Fig. 10B). Based on the simulations, 37 sites had not yet reached the DL of maximum interface in 1985, whereas two were at maximum-interface DL, and 31 were beyond it (Fig. 10C).

Aggregated Interface Length versus Year

Aggregated 1985 simulation interface was 406,051 pixel-lengths (12,182 km)--82% of the aggregated maximum interface of 496,969 pixel-lengths (14,909 km). According to our estimates from Wicker's (1980) map data, the average annual change in DL in the USGS-topographic-map areas of our study sites varied from 0.125% to 1.145% per year (Appendix H). Using these trends and the year of our TM image to relate DL to time, we transformed our individual-site plots of interface versus DL to the lobe-aggregated plots of interface versus time in Figure 11. Our hindcasts (1956-1985) and predictions (1985+) of interface are plotted as fractions of total maximum interface. The interface curves do not reach 1.0 because all the simulated marshes will not reach their interface maxima concurrently. The 1985 points are near the two maxima on the ascending side. These results suggest that total land-water interface in both deltaic areas has been increasing, but will soon begin decreasing. If the estimated linear

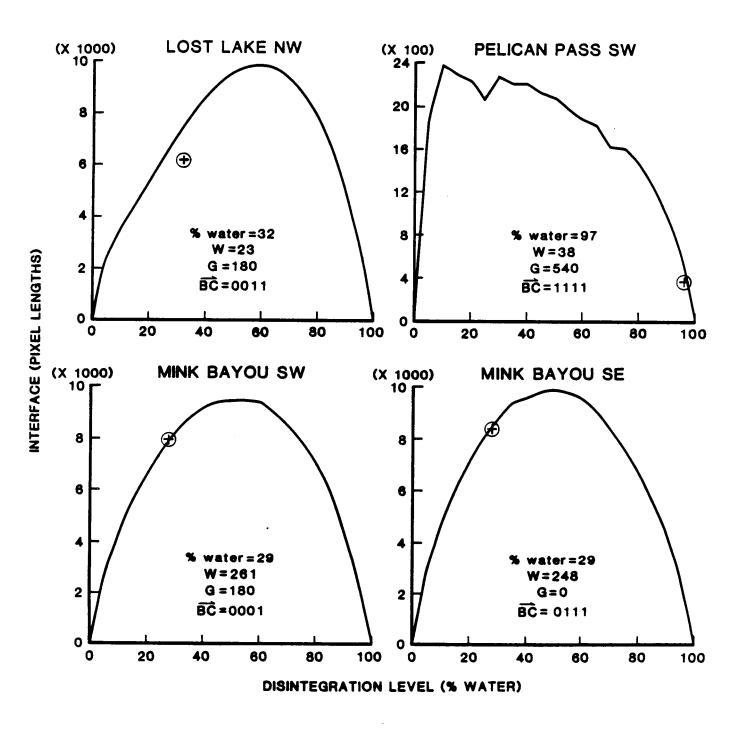


Figure 9. Simulated interface versus disintegration level (DL) for 4 of the 70 cases showing the modeling coefficients (W, G, and BC) used in the simulation.

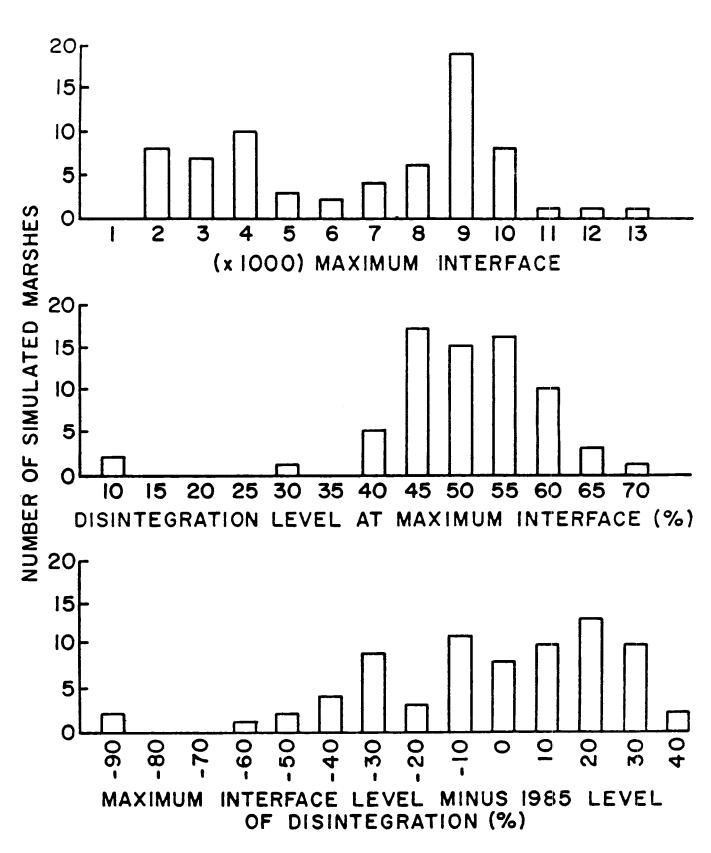
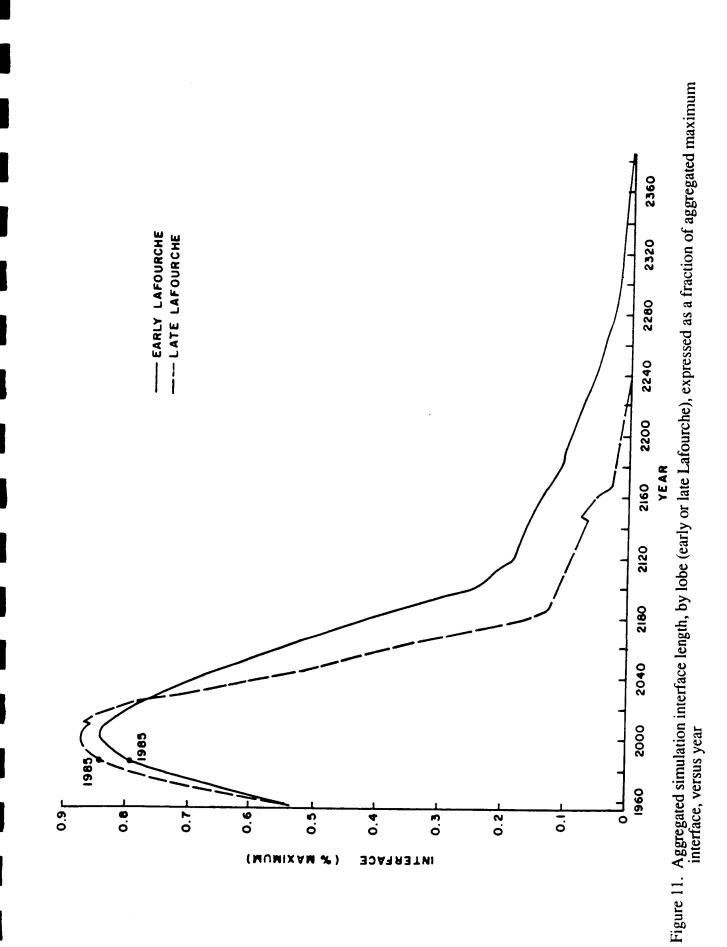


Figure 10. Frequency distributions of cases by simulation maximum interface (A), disintegration level (DL) at maximum interface (B), and maximum-interface DL minus 1985 DL (C). Numbers on the abscissas are the upper values of each interval.



trends in DL are accurate, the decline will begin about 1995. Sasser et al. (1986) reported an exponential rate of loss in our study area. On the other hand, recent observations by Liebowitz (personal communication) suggest that some large water areas that appeared between 1956 and 1978 may not be growing.

Possible Impacts on Fisheries

We found a statistically significant (p < .0001) relationship between brown shrimp catch and interface. The equation is as follows:

 $Y_p = -61.046 + 277.55 \cdot X_I - 0.39198 \cdot X_2 - 0.12948 \cdot X_3$ (3) where $Y_p =$ catch per unit area, $X_I =$ interface length in pixel sides, $X_2 =$ rainfall in centimeters, and $X_3 =$ hours from April 9 through 30 in which temperatures were below 20°C. The equation explained 49% of the variation in catch for 1960-1987 (Appendix I). Interface length alone explained 32%. The percentage variation in annual catch explained by the equation was high, considering that effort, usually a major factor influencing catch, was not included as an independent variable. Using our interface projections and assuming average conditions of the other independent variables, the equation predicts that brown shrimp catches dependent upon Barataria, Timbalier, and Terrebonne bays may fall to zero within 75 yr (equation confidence limits 52 and 105 yr). Confidence limits do not include the error associated with predicted interface.

Discussion

Our model and expert system appear to have been successful in simulating general features of the spatial patterns of most study sites. The model was not designed to reproduce the exact locations of land and water in each study site but, rather, the general characteristics of the landwater pattern. Since the model is probabilistic, it produced a different pattern in every execution with the same W-G-BC combination. Necessary built-in restrictions such as having only six possible values of G and the same G for all water borders limited our versatility in matching spatial patterns. We could have matched the interface length of the TM scenes more closely had we not also matched the side-adjacency statistics. But selecting W-G-BC on the basis of both interface length and the side-adjacency statistics increased the probability that the trajectory of interface change with land loss during each simulation of marsh disintegration was realistic for the site.

The statistically significant fit of a parabola to the plot of interface-length versus DL of our 70 TM scenes (Fig. 4) supports Browder et al.'s (1984) first conclusion: In the progress of marsh decay, interface length increases initially, reaches a maximum, and then decreases. The scatter of points about the parabola (Fig. 4) supports their second conclusion: The magnitude of maximum interface (and, consequently, the trajectory of interface change with land loss) differs from marsh to marsh. Our expert system selected model-parameter values to simulate marsh disintegration at each site on the basis of spatial-pattern indices measured in the site's TM scene. The site-specific parameters produced considerably different trajectories of interface change with land loss. Maximum interface varied from about 2,000 pixel-lengths (60 km) to over 13,000 (390 km) (Fig. 10A), and DL at maximum interface varied from about 10% to 70% (Fig. 10B) in the 70 simulations. Simulations with the Browder et al. (1984) model consistently reached maximum interface at a DL of about 50%. Apparently, the greater power of the G-weighting coefficient in our model gave it more flexibility in simulating interface trajectories. Nevertheless, our 70 simulation results were centered around a mean DL at maximum interface of 52.7% (S.D. = 9.95).

In the plot of interface length versus DL with TM-scene data, DL at maximum interface was between 30% and 50%. This might appear to conflict with the site-specific simulation results, summarized in Figure 10B, which suggest that interfaces reach their maxima in most of the sites when they are 45%-60% disintegrated. But, when we plotted the simulated 1985 interface lengths against DL, we found that these data, too, reached maximum interface at 30%-50% DL. Apparently, a plot of interface length versus DL at many sites in different stages of disintegration does not precisely reflect the generalized shape of the curve of interface versus DL at the individual sites.

The resolution of TM imagery seemed adequate for this analysis. Many water features were recognizable that would not have been noticeable in MSS imagery. Our model might be useful for roughly estimating the history of interface length with disintegration in other marshes, even in the absence of the detailed spatial data we acquired for our 70 sites. On the basis of our results, model coefficients could be set as follows: BC = observed water borders, G = our

mean or modal value for salt or brackish marsh (Fig. 7A), and W = our mean or modal value for that border class (Fig. 7B).

Summary

We demonstrated with TM imagery the general relationship of land-water interface length to stage of disintegration. We then simulated the disintegration of 70 specific marshes from hypothetical starting points of solid land, through their present states of disintegration, to total conversion to water. We used unpublished data from digitized maps (Wicker 1980) to quantify site-specific disintegration rates, and we hindcasted and forecasted land-water interface as a function of time. We then aggregated the site-specific data to produce an estimate of interface length, by year, on each lobe. The relationship with time may be tenuous because we assumed a linear trend based on only two points in time. Nevertheless, relating our results to time, even if only roughly, helps the reader comprehend the immediacy of the problem.

We found a statistically significant relationship between a time series of fishery catch data and the length of the land-water interface. Others have found relationships between spatial data on fishery catch and interface. Our analysis may overestimate the importance of interface to brown shrimp production because the conversion of freshwater marsh to brackish marsh (or some other factor not included in our equation) might ameliorate the effect of interface loss in salt and brackish marsh. Nevertheless, the shape of our curve of interface over time, today's location on that curve, and our contribution to the mounting evidence relating fishery catches to interface length should be seriously regarded.

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References

- Barrett, B. B., and M. C. Gillespie. 1975. 1975 Environmental conditions relative to shrimp production in coastal Louisiana. Louisiana Wildlife and Fisheries Tech. Bull. 15.
- Baumann, R. H., J. W. Day, Jr., and C. A. Miller. 1984. Mississippi deltaic wetland survival: sedimentation vs coastal submergence. Science 224:1093-1095.
- Boesch, D. F., and R. E. Turner. 1984. Dependence of fishery species on salt marshes: the role of food and refuge. Estuaries 7:460-468.
- Browder, J. A., H. A. Bartley, and K. S. Davis. 1984. A probabilistic model of the relationship between marshland-water interface and marsh disintegration. Ecological Modeling 29:245-260.
- Chabreck, R. H. 1972. Vegetation, water and soil characteristics of the Louisiana coastal region (map). Louisiana State University, Agricultural Experiment Station Bull. 664, Baton Rouge.
- Chabreck, R. H., and G. Linscombe. 1978. Vegetative type map of the Louisiana coastal marshes. Louisiana Department of Wildlife and Fisheries, New Orleans.
- Dow, D. D. 1982. Software programs to measure interface complexity with remote-sensing data, with an example of a marine ecosystem application. NASA Report No. 219. NASA Earth Resources Laboratory, NSTL, Miss.
- Dow, D. D., and R. W. Pearson. 1982. SLIN: a software program to measure interface length, NASA Report No. 208. NASA Earth Resources Laboratory, NSTL, Miss.
- Faller, K. H. 1979. Shoreline as a controlling factor in commercial shrimp production. NASA Tech. Memo. 72-732. Earth Resources Laboratory, National Space Technologies Laboratory, NSTL, Miss.
- Gagliano, S. M., K. J. Meyer-Arendt, and K. M. Wicker. 1981. Land loss in the Mississippi River deltaic plain, Transactions of the Gulf Coast Association of Geological Societies 31:295-300.
- Gosselink, J. G. 1984. The ecology of the delta marshes of coastal Louisiana: a community profile. FWS/OBS-84/09. Office of Biological Services, U.S. Fish and Wildlife Service, Slidell, La.

- Graham, M. H., B. G. Junkin, M. T. Kalcic, R. W. Pearson, and B. R. Seyfarth. 1984. ELAS: Earth Resources Laboratory Applications Software. Vol. 2, ELAS User's Guide. NASA Earth Resources Laboratory, NSTL, Miss.
- Kesel, R. H. 1988. The decline in the suspended load of the lower Mississippi River and its influence on adjacent wetlands. Environ. Geol. Water Sci. 11:271-281.
- Leibowitz, S. G., and J. M. Hill. 1988. Spatial analysis of Louisiana coastal land loss, In R. E. Turner and D. R. Cahoon, eds., Causes of Wetland Loss in the Coastal Central Gulf of Mexico, Vol. 2. OCS Study/MMS 87-0120. Minerals Management Service, New Orleans.
- Mandelbrot, B. B. 1967. How long is the coast of Britain? Statistical self similarity and fractional dimension. Science 56:636-638.
- Moran, P. A. P. 1948. The interpretation of statistical maps. Journal of the Royal Statistical Society, Series B, 10:243-251.
- Nichols, S. 1984. Updated assessments of brown, white, and pink shrimp in the U.S. Gulf of Mexico. Stock Assessment Workshop 1984, Southeast Fisheries Center, National Marine Fisheries Service, NOAA, Miami, Fla.
- Pun, Y. T. 1981. Entropic thresholding, a new approach. Computer Graphics and Image Processing 16:210-239.
- Sasser, C. E., M. D. Dozier, J. G. Gosselink, and J. M. Hill. 1986. Spatial and temporal changes in Louisiana's Barataria Basin Marshes, 1945-1980. Environmental Management 10:671-680.
- Turner, R. E., and D. R. Cahoon. 1987. Causes of wetland loss in the coastal central Gulf of Mexico. 3 vols. OCS Study/MMS 87-0119, 120, 121. Minerals Management Service, New Orleans.
- Upton, G., and B. Fingleton. 1985. Spatial Data Analysis by Example. Vol. 1. John Wiley, New York.
- Wicker, K.M. 1980. Mississippi deltaic plain region ecological characterization: a habitat mapping study. FWS/OBS-79/07. U.S. Fish and Wildlife Service, Slidell, La., and Minerals Management Service, New Orleans.
- Zimmerman, R. J., T. J. Minello, and G. Zamora, Jr. 1984. Selection of vegetated habitat by brown shrimp, *Penaeus aztecus*, in a Galveston Bay salt marsh. U.S. Fisheries Bull. 82:325-336.

Appendix A

Background and Methods

Louisiana's Coastal Wetlands: Geological Background and Previous Remote Sensing Studies

Methods Overview Model Specification Model Expansion, Refinement, and Sensitivity Testing Study Site Selection Image Processing and Analysis Measurement of Spatial-Pattern Statistics Description of the Expert System

References

- Table A1. Look-up table used to classify water and land identified by the ELAS shoreline length module into water pixels and land pixels with zero, one, two, three, or four sides adjacent to water.
- Table A2. Calculation of weighted mean W, DE, and CV from a G-BC look-up table in which G = 0, BC = 0000, and DL = 10.90%.
- Figure A1. The maximum extent of the influence of deltaic lobes of the Mississippi River on the present geomorphology of Louisiana's coastal wetlands.

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Louisiana's Coastal Wetlands Geological Background and Previous Remote-Sensing Studies

The continuing disintegration of the coastal marshes of Louisiana is one of the major environmental problems of the nation. The average rate of loss for the last 20 yr has been approximately 104 km²/yr (Gagliano et al. 1981). At this rate, Louisiana's coastal marshes will be gone in 145 yr. Prevailing evidence suggests that the marsh disintegration results from local imbalances between building processes, such as sedimentation and the growth and accumulation of dead vegetative matter, and destructive processes, such as sea level rise, crustal subsidence, erosion, and compaction (Gosselink 1984). Local elevation gradients within the marsh are so low that small changes in water level or land elevation can cause large changes in land and water area (Sasser 1977; Baumann 1980). Water management structures, navigation cuts and channels, and other alterations by man appear to accelerate the disintegration rate (Johnson and Gosselink 1982; Dozier 1983; Gosselink 1984; Turner et al. 1984).

The problem of marsh loss in Louisiana is relevant to fishery management because Louisiana leads the nation in landings of fishery products, and most of the landed species are dependent upon estuaries and their associated tidal marshes. Coastal marshes contribute to estuarine food chains through the export of organic detritus, and the shallow, protected water of marshes serves as fish and shellfish nursery grounds, promoting survival and growth of the young.

Remote-sensing studies by Faller (1979), Dow (1982), and Gosselink (1984) suggest that the abundance of fishery species is more strongly correlated with the length of the interface between land and water in the marsh (shoreline) than with actual area of marshland. Observations from a field study by Zimmerman et al. (1984) support this conclusion. Simulations from a theoretical computer model by Browder et al. (1984) suggested that landwater interface initially increases with marsh disintegration but reaches a maximum when the marsh is 50% water and decreases thereafter. The degree of change in interface with each incremental loss of marsh land and the maximum length of interface attained are a function of the order in which segments of land are converted to water and the resultant pattern of

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distribution of land and water. The more clustered the segments of land converted to water, the lower the rate of change and less the maximum interface.

In evaluating the potential effect of marshland loss on fisheries, the first two critical factors to consider are (1) whether land-water interface in actual disintegrating marshes is currently increasing or decreasing, and (2) the magnitude of the change.

This study used Landsat Thematic Mapper (TM) data covering specific sample marshes in coastal Louisiana to (1) test conclusions from the Browder et al. (1984) model with regard to the stage in disintegration at which maximum interface occurs; (2) further explore the relationship between maximum interface and the pattern of destruction of land and water suggested by the model; and (3) determine the direction and degree of change in land-water interface in relation to land loss in actual marshes.

Louisiana's coastal marshes were ideally suited for this examination for several reasons. First, the large, contiguous expanses of marsh enabled us to sample large areas containing only wetlands. Second, this region has been the subject of many scientific investigations concerning ecological principles, geologic processes, and experimental use of remote-sensing techniques. Third, geologic changes are occurring very rapidly here, and fourth, Louisiana's coastal marshes are the most extensive in the United States and support a high proportion of the total U.S. production of estuarine-dependent fish and shellfish.

The coastal wetlands of Louisiana were formed as deltas of the Mississippi River and its tributaries. The large, heterogeneous expanse of deltaic wetlands along the Louisiana coast is very young geologically. This area was formed within the last 3,000-5,000 yr as a series of overlapping deltaic lobes of differing ages (Fig. A1). Instability is a characteristic of youthful geologic environments. Subsidence, a complex set of processes, has pronounced effects on wetlands near sea level. Isostatic adjustments in the form of crustal downwarping from sedimentary loading; tectonic processes that occur contemporaneously, such as folding, fracturing, flowing, and growth faulting; consolidation of underlying sediments due to the weight of natural features (e.g., natural levees); and differential compaction related to textural variability are among those natural processes involved in submerging this coastline. Human activities in the form of fluid withdrawals (hydrocarbons and water), marsh dewatering through reclamation processes, and sediment consolidation resulting from building structures on wetlands all exacerbate coastal submergence. The above subsidence factors, combined with eustatic sea

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level rise, have given coastal Louisiana the fastest-submerging coastline in the United States (Hicks 1981).

Submergence results in the "drowning" of marshes and their conversion to bay and lake environments. Sedimentation can balance the effects of submergence and, via the Mississippi River and its tributaries, has resulted in development of the Mississippi delta lobe. The geologic record indicates that, on the average, a major delta lobe complex will build and enlarge over approximately 1,000 yr. This period is followed by one of abandonment characterized by wetland loss, which also lasts approximately 1,000 yr. As one delta complex is being abandoned, another one is simultaneously building. Throughout at least the Holocene, the Mississippi deltaic plain has always concurrently had areas of development and abandonment. Presently, however, the leveeing of the Mississippi River and maintenance of its present course, combined with reductions in sediment loads (Tuttle and Combe 1981) and debouchment of sediment at the edge of the continental shelf, have resulted in widespread wetland loss. The construction of ship channels, pipeline canals, and access canals for hydrocarbon exploration and production has both contributed to and accelerated these losses. Acceleration occurs through the effect of these structures on salinity distributions and sediment deposition. For instance, canals promote saltwater intrusion, which results in the death of brackish-water marsh vegetation and thus retards the accumulation of organic soils. Spoil banks associated with canals prevent deposition of sediment on the marsh surface and reduce exchanges of water and materials between the marsh and open water. The natural geological process of wetland deterioration, which would otherwise take place over several centuries, appears now to have been compressed into several decades.

Chabreck (1972) distinguished four major types of Louisiana coastal marshes on the basis of vegetation: fresh, intermediate, brackish, and saline. Several investigators have found significant differences between these marsh types in various soil, water quality, and other parameters, thereby supporting Chabreck's classification. Gosselink et al. (1979) found considerable differences in the length of land-water interface per unit area among the four marsh types in the neighboring Chenier Plain (marginal Mississippi deltaic plain) of southeast Texas and southwest Louisiana.

Sasser et al. (1986) used photo interpretation of aerial photographs in combination with a computer-based geographic information system (GIS) to detect changes in the percentage water

within wetlands on the late Lafourche delta lobe. They found a pattern of general degradation in wetland area: marshes were degrading into various densities of shallow water bodies. Of the marsh and natural levee area, 91% was solid or less than 10% water in 1945. By 1956, only 77% of the marsh was less than 10% water; by 1969, only 46% was; and by 1980, only 28%. They noted two patterns of disintegration. In one, small, randomly spaced water bodies developed within solid marshes and gradually grew into larger water bodies. In the other, land was lost along the margins of major water masses, as if by mechanical wave attack or erosion. The first pattern seemed to be the more important.

Rosen (1980), in his study of Chesapeake Bay, concluded that shorelines with low tidal ranges have higher rates of erosion than areas with higher tidal ranges, possibly because higher tidal ranges form beaches of higher elevation. On these beaches, storm surges are less likely to reach the elevation of fastland (bluff or dune) material to augment erosion, and wave energy is distributed over a greater distance in the course of a tidal cycle. The tidal range in Chesapeake Bay varies from 0.36 to 1 m over a distance of 120 km. The tidal range in the north-central Gulf of Mexico is approximately 0.6 m.

Leibowitz and Hill (1988) used digital habitat maps for 1956 and 1978 from the U.S. Fish and Wildlife Service (Wicker 1980) to quantify change in coastal marshes during the 22-yr period and to evaluate various possible causes of the change. Their study covered our two study areas--the late Lafourche lobe and the early Lafourche lobe (referred to as Terrebonne in their study). Water, wetland, and upland could be distinguished in the data, which were classified according to the Cowardin et al. (1979) system. Boundaries between saline and freshwater zones were also defined on the basis of vegetation. Liebowitz and Hill classified each map cell on the basis of a comparison of 1956 and 1978 habitat maps as follows: areas that were fresh in 1956, but saline in 1978; areas that changed from saline to fresh between 1956 and 1978; and areas that remained saline during the 22 yr. They also identified the cells in each habitat category that changed from land to water during the 22 yr. Their results revealed a 37% net area change from salt to fresh on the late Lafourche lobe and a 16% net area change from fresh to salt on the early Lafourche lobe. The highest rate of land loss on the late Lafourche lobe was 27% and occurred in the fresh-to-salt area. The highest rate of land loss on the early Lafourche lobe was 16% and occurred in the fresh-to-fresh area. By statistical comparisons, they ruled out saltwater intrusion as a reason for land loss on the early

Lafourche lobe, but concluded that it could be a cause of land loss on the late Lafourche lobe. The highest rates--47%-55%--occurred in the mud flat and beach/dune/reef habitats. Loss rates in fresh and saline marsh averaged approximately 18%. Loss from shoreline erosion accounted for only 2.1% (early Lafourche) and 3.2% (late Lafourche) of all land loss. Thus, the major form of land loss for all three regions was the conversion of land to inland open water (lakes, ponds, or bays).

Several studies have used TM and Landsat MSS imagery with collateral data, such as fish abundance and vegetative biomass, to examine the role of coastal wetlands in estuarine food chains and the production of estuarine-dependent fish and shellfish. These studies were supported by the development of software routines used to determine shoreline density (Faller 1977) and shoreline length (Faller 1977; Dow and Pearson 1982), to identify water bodies (Butera 1982a), and to measure the distance between land-cover classes (Butera 1982b). Faller (1979) found a strong correlation between shrimp yields and shoreline density in subareas of the Louisiana coastal zone. Dow (1982) expanded Faller's (1979) approach and developed predictive equations that related the abundances of selected species of fish and shellfish to shoreline-length estimates for subareas of Apalachicola Bay, Florida. The findings of both authors suggest that abundances of some fish and shellfish could be influenced by the density and length of the marshland-water interface. Butera and Seyfarth (1981) and Butera et al. (1984) used water-body identification, distance measures, shoreline density, and vegetative biomass estimates to quantify organic carbon export into nearby water bodies.

Methods

Overview

We expanded an existing model (Browder et al. 1984) so that it could simulate marshes of substantial size, used actual marshes to calibrate the weighting factors of the models, and then used the model to simulate the disintegration over time of each sample marsh. Model calibration was accomplished by quantifying the spatial-pattern statistics of the sample marshes and matching them to the spatial-pattern statistics expected from simulated marshes, based on a series of simulations in which W, G, and the number of water borders (BC) were varied.

The process consisted of nine steps: (1) expansion, refinement, and sensitivity testing of the model; (2) selection of sample sites; (3) analysis of imagery; (4) measurement of spatial-pattern statistics; (5) development of a knowledge base and an expert system; (6) calibration of the model to the sample marshes; (7) simulation of the disintegration patterns of the sample marshes; (8) evaluation of simulation results; and (9) interpretation.

Model specification. The model used in this study is the second generation of a stochastic spatial computer model introduced by Browder et al. (1984). In the initialization of the model, marsh dimensions are defined in terms of the numbers of rows and columns of pixels. Each pixel can exist in one of two states, land (emergent vegetation) or water. Initially, all the pixels are land and the marsh is solid. One land pixel is converted to water at each iteration. The actual pixel converted is determined by a random number generator linked to a probability function that incorporates two weighting factors that approximate the natural processes of interior marsh decay (the W factor) and shoreline erosion (the G factor). The W factor determines disintegration probability on the basis of the number of sides that the pixel is bordered by water. The G factor governs the probability that the pixel will disintegrate if it borders the main water body. The probability weight of each pixel is calculated by the equation:

 $F_{i,j,k} = 1 + W S_{i,j,k} + G_I B_{Ii,j} + G_2 B_{I2i,j} + G_3 B_{3i,j} + G_4 B_{4i,j}$ (1) where W = weight coefficient for each side adjacent to water; S = number of sides adjacent to water; G = weight coefficient for pixels adjacent to a major outside water body; and B = a Boolean value (1 or 0) indicating whether the pixel is adjacent to a major outside water body. The probability weight of a given pixel changes throughout the simulation, depending on what happens to other pixels, particularly those adjacent to it.

In having the weighted probability function approximate natural processes of interior marsh decay (W) and erosion due to tidal action or wind-induced turbulence along the edge of major water bodies (G), we did not assume that marsh loss is a random process, but merely that it could be simulated by a weighted, randomly driven function.

The model simulates the entire process of disintegration, starting with solid land and ending with solid water. Each iteration represents the passage of time, although the units of time are not specified.

At each iteration of the simulation, a counter keeps track of the percentage area represented as water, referred to throughout this discussion as the "level of disintegration," and the length of the land-water interface. The latter is expressed in terms of pixel-lengths, the length of one side of the square pixel; therefore, measuring interface length consisted of counting the number of "joins" between land pixels and water pixels. Thus, interface, as we measured it, is exactly homologous to the "black-white join" (J), the spatial autocorrelation parameter that Moran (1948) introduced into the literature of quantitative geography. Upton and Fingleton (1985) described the common relationship between the join statistic and other spatial-pattern parameters, such as that of Cliff and Ord (1973), and defined the cross-product statistic, R, which is equal to $2 \times J$.

Upton and Fingleton (1986) provide an intricate set of equations for calculating R, the expected value of R (E[R]), and the variance of the expected value. E(R) assumes a random distribution of black and white (or land and water) cells. R departs from E(R) to the extent that like cells are clumped (R < E[R]) or uniformly distributed (R > E[R]). They provide simpler equations for calculating J, E(J), and var E(J) for cases in which the area is regular-sided and square in configuration (their equations for the R statistics are more general). In our simulations, we were able to determine J simply by keeping a running total of the number of land-water joins created at each conversion of a land pixel to a water pixel. A method related to counting was used to determine the number of land-water joins in satellite images classified as land and water. Our observations indicate that, for a square area with regular sides, E(J) is approximately equal to one-half the number of land-water joins of an area of the same

dimensions having a checkerboard pattern of distribution of land and water. This can be calculated as follows:

$$E(J) = 2 N^2 - 4 N$$
 (2)

where N = the number of rows = the number of columns.

The weighting factors affect the order of disintegration of marsh land pixels and the resultant distribution of land and water in the simulated marsh. The higher the values of the weighting factors, the more clumped the water pixels. By affecting the spatial distribution of water pixels, the weighting factors determine interface length in simulated marshes. Taking advantage of this relationship, the approach we took to simulating the disintegration of actual marshes was to use spatial pattern, as expressed by level of disintegration, interface length, and other spatial pattern statistics of the actual marshes, compared to those from simulated marshes, to select W and G weighting factors for the model. The other spatial-patterns statistics that were used were numbers of water pixels with zero, one, two, three, or four sides adjacent to other water pixels, and numbers of water pixels on each of the marsh's four borders. The distribution of water pixels by size of water clusters at the current (i.e., December 1984) level of disintegration was used to test the fit of the simulated marsh to the actual marsh. Comparison of simulated marshes to actual marshes in general suggests that the function will work well for simulating reticulated marshes, such as those on the Gulf coast, although it might not work well for marshes with a more dendritic pattern of land and water, such as those along the U.S. Atlantic coast.

Model Expansion, Refinement, and Sensitivity Testing

The first phase in the study was improving the model. Our improvements were guided by a series of sensitivity tests: (1) tests of the effects of the W and G weighting factors, varied separately; (2) tests of the effect of marsh geometry (i.e., length, relative to width); and (3) tests of the effect of marsh size, in terms of number of pixels.

In the original version of the model, only the pixels *initially* on the major outside water body had the G weighting (B = 1). The G effect was inconsequential in sensitivity tests with the original model, particularly as the size of the marsh simulated was increased. On the basis of this observation, the model was revised so that any pixel, regardless of original location, could eventually be assigned B = 1. The G factor in the present version of the model has a much greater effect than that in the earlier version.

Other sensitivity tests indicated that the geometry of the marsh (i.e., ratio of length to width) affected the trajectory of change in interface relative to W and G and greatly complicated the process of examining interface as a function of W and G and the number of water borders to the marsh (i.e., simulation results differed depending upon whether a water border was the long or the short border). We decided to work with square marshes, both simulated and actual, in order to avoid this complication.

To eliminate another complicating variable--scaling--we decided to simulate marshes of the same size (same number of pixels) as our sample sites. We determined that it would be practical to simulate marshes up to 192×192 pixels, although not with replication. A site represented by 192×192 pixels covers 33.18 km^2 and is approximately one quarter of the area covered by a 7.5-minute U.S. Geological Survey topographic map.

Increasing the size of the simulated marsh necessitated streamlining the algorithm for weighting disintegration probability and converting land pixels to water pixels. In the original algorithm, each pixel, identified by its x,y coordinate, was repeated on the list the same number of times as its probability factor (F in equation 1). Each item on the list had a unique number, and the pixel selected was the one that corresponded to the random number at that iteration, providing that it had not already been converted to water at a previous iteration. All occurrences of pixels that had been newly converted to water were cleared from the list at five periodic intervals throughout the simulation. The process got slower and slower as the need for purging the list approached. This algorithm was too slow and awkward to be scaled up in the same form. In our revision, each pixel appears on the numbered list only once, but its probability factor is listed with it. Two random numbers are associated with each selection. The first random number makes a tentative selection, and the second determines whether the pixel is eligible. Eligibility depends on whether the pixels's probability factor is larger than the random number. The selection process continues, with two new random numbers generated each time, until the selection of an eligible pixel is made. Of course, the first random number-the one that makes the tentative selection--is a uniform random number from 0 to 1 that is multiplied by the largest probability factor on the list. A pointer system keeps track of the

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pixels on the list and eliminates from the list the pixel that has been converted to water at each iteration.

The model and all ancillary programs were written in C and were executed on an AT&T PC-7300, a 16-bit computer that has a Unix-V operating system.

Study Site Selection

The study sites are located in salt and brackish marsh areas on two abandoned delta lobes of the Mississippi River, the early lafourche and the late Lafourche. The early Lafourche lobe was an actively prograding lobe within the last 1,800 yr; the late Lafourche lobe was active as a main distributary of the river within the last 600 yr. On each lobe we selected sites that corresponded to the boundaries of five contiguous U.S. Geological Survey 7.5-minute topographic maps. Areas defined by each topographic map were divided into four contiguous quarters, each encompassing an area 192 elements wide and 192 scan lines long on the TM image. The intersection of the four quarters was aligned to correspond to the center point of each topographic map. Each area corresponding to a quarter area of the 10 topographic maps was a potential sample site. After excluding sites with upland vegetation and sites for which no cloud-free TM images were available, we had 72 samples to use in the study: 40 salt marsh sites (20 on each lobe) and 32 brackish marsh sites (19 on the early Lafourche lobe and 13 on the late Lafourche lobe). Salt and brackish marshes were distinguished by means of the U.S. Fish and Wildlife Service habitat maps (Cowardin et al. 1979).

Because of small errors in TM imagery, pixels are neither exactly square nor exactly the same size; therefore, it was necessary to eliminate several pixels on the outer boundaries of imagery corresponding to each topographic map in order to have a 192 x 192 image. Our sample images therefore do not provide complete coverage of the area--small strips at the boundaries of the topographic maps are missing. Selecting square samples (samples having the same number of rows and columns of pixels) greatly simplified the analyses of this study in several ways. First, we had fewer alternatives to consider in sensitivity analysis and constructing look-up tables. Second, we could use simpler and less time-consuming equations for estimating spatial autocorrelation statistics. The quarter was the largest square unit into which a topographic map could be evenly divided that could be simulated with practicality in the same dimensions by our computer model on available dedicated hardware.

Image Processing and Analysis

TM scenes were analyzed on the Fisheries Image Processing System (FIPS) maintained by NMFS in Slidell, Louisiana. FIPS uses a Sperry-Univac V77/600 mini-computer, color image display device, and other hardware to process remotely sensed digital imagery. The software is a modified version of the Earth Resources Laboratory Applications Software (ELAS) (Graham et al. 1984).

The TM image acquired for the project represented one of the few relatively cloud-free images covering southern Louisiana (quads 1 and 2 in path 22 and row 40 of the World-Wide Reference System). The Landsat overflight occurred on 2 December 1984 (Scene ID: 50276-16022) and covers most of the Mississippi deltaic plain.

TM images of the sites were georeferenced to fit a Universal Transverse Mercator projection with a north-south orientation. The ELAS modules PMGC and PMGE (Graham et al. 1984) were used to accumulate ground control points, generate polynomial least-squares mapping equations, and resample the image using the bilinear interpolation technique. The average registration accuracies ranged from 22 to 56 m.

Land and water were distinguished in the TM images by first generating a product image from bands 4 and 5 and then applying the global thresholding technique developed by Pun (1981).

Measurement of Spatial-Pattern Statistics

We generated 72 binary land-water images from the product images of the salt and brackish marsh sites. Sequential ELAS commands set up for batch processing were used to measure the following spatial-pattern parameters in each image: (1) total numbers of land and water pixels; (2) total numbers of water pixels by scan line and by element column; (3) the length of the land-water interface, expressed as the total number of land-water joins; (4) total numbers of water pixels with sides adjacent to zero, one, two, three, and four other water pixels; and (5) water-body size frequencies. In determining the total number of water pixels with sides adjacent to other water pixels, we excluded the pixels at the boundary of the sample to avoid biasing the distribution of pixels toward those having less than four sides adjacent to water.

The total number of land-water joins in each image was tabulated using a three-step process. First, an intermediate image was generated using the ELAS shoreline-length (SLIN) module (Graham et al. 1984). SLIN uses a 3-x-3 moving window technique to classify each land pixel adjacent to water into one of 69 shoreline categories (Dow 1982; Dow and Pearson 1982). Second, we used a look-up table (Table A1) to convert the SLIN image into an image file comprising six classes: (1) land; (2) water; and (3) shoreline pixels with one, two, three, or four sides adjacent to water. Finally, we determined the total number of land-water joins in each sample site by enumerating the number of land pixels sides bordering water pixel sides.

The total number of water pixel sides adjacent to other water pixels was tabulated using a modification of the technique used to count land-water joins. Two changes in the processing sequence were required: (1) water pixels adjacent to land were defined as shoreline pixels during processing with the SLIN module and (2) an additional processing step with a new look-up table was required to correctly classify water pixels with zero, one, two, three, or four sides adjacent to other water pixels.

As Hutchinson (1957) originally pointed out and first Richardson (1961) and then Mandelbrot (1967) elaborated upon, the length of an irregular shoreline is, to some extent, a function of measurement unit. Our measurements of land-water joins and, possibly, the other spatial-pattern statistics, are valid only at the resolution of the TM imagery, the 30-x-30-m pixel. Future measurements cannot be compared to ours unless the same measurement unit is used.

Description of the Expert System

An expert system was developed to select the model parameters--W, G, and BC--that would best approximate the spatial patterns of each study site. The expert system consisted of a knowledge base and a decision process. The knowledge base indicated how each of the spatial pattern indices--interface length and the four side-adjacency statistics--varied as functions of W, G, and BC (border condition). The decision process consisted of the rules for selecting the best W-G-BC combination.

To build the knowledge base for the expert system, we ran simulations with all possible W and G combinations from the set [0, 4, 20, 60, 180, and 540] for six types of study-site border conditions: 0 = no water border, 1 = 1 water border, 2 = 2 adjacent water borders, 3 = 2 opposite water borders, 4 = 3 water borders, and 5 = 4 water borders. For border condition 0, the set was extended to include W = 1620 and 9720. Each simulation contributed information to 21 tables. Each table contained interface length and side-adjacency information collected at

a 5% increment of DL (level of disintegration, or water area as percentage total area). We compiled 21 tables (one for each increment of DL) for each value of G and for each border condition (a total of $6 \times 5 = 30$ sets of 21 tables). For border condition 0 (no water border), only one set of 21 tables was compiled, since G must equal zero. For each of the other data sets, there were 21 tables for each G value.

The following statistics from each study site were used in the decision process: DL; interface length; and Adj-0, Adj-1, Adj-2, Adj-3, and Adj-4 (number of water pixels having 0, 1, 2, 3, or 4 sides adjacent to water). Target BC was an additional factor in the decision process.

The DL of the study site was used to determine which tables were accessed. The tables of the nearest DL's on either side of the study-site DL were accessed. For instance, if the DL of the image was 32%, then the tables for DL's 30% and 35% were accessed. Interpolation between levels was then used to produce, for every G value and border condition, a table of values of spatial-pattern indices for each of the six values of W for the specific DL of the study site.

Then, for each G value and border condition, the study-site interface and side-adjacency values were compared with values for the spatial-pattern index in the tables prepared for the specific disintegration level. If the study-site value for a spatial-pattern index was within the range of values for that index on a particular table, exact matching or interpolation between values was used to estimate W on the basis of that index, given the G value and border condition of that table. If the value of a given index from the study site was not within the range of values for that index in a table, then W could not be estimated from that particular index and table.

We usually obtained several estimates of W from a given G-BC table. A weighted mean W for the specific G-value and border condition was obtained from these. In cases where a parabolic relationship between the parameter and W occurred, more than one estimate of W was sometimes obtained for the same index and table. In such cases, each estimate was used alternatively in calculating a weighted mean until we had calculated all possible weighted means from the indices. For instance, interface might yield W = 2, 4; Adj-0, W = 180, 193; and Adj-3, W = 300. Then 2 x 2 x 1 weighted mean W's were calculated. One would involve 2, 180, and 300; another 2, 193, 300; another 4, 180, 300; and another 4, 193, 300. Weighting was a function of the number of water pixels involved in each parameter estimate of W.

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the parameter was the estimate of the number of pixels involved in the estimate of W from that spatial-pattern index. Weighted mean W's were calculated as follows:

Weighted Mean W = Sum
$$(W_i V_i) / Sum (V_i)$$
 (3)

where W_i = the estimate of W from statistic i, and V_i = the number of pixels involved (index value), statistic i. Only the water pixels of the spatial-pattern indices involved in the specific calculation of the weighted mean W were summed. As mentioned above, if the index value from the sample was not within the range of values for that index in a particular table, an estimate of W based on that index could not be obtained.

The coefficient of variation (CV) of each weighted mean W also was calculated, as follows:

$$CV = (Variance)^{1/2} / Weighted Mean W$$
 (4)

In addition, the sum of the water pixels used in calculating the weighted mean W was retained as a "decision number" (DE) for later use in the selection process. Table A2 lists the weights, the W's, and the calculations of weighted mean W, DE, and CV.

By the above procedure, the expert system estimated many W-G-BC combinations for each study site. Weighted mean W's, CV's, DE's, and their corresponding G's and BC's were stored in solution files specific to each study site. The file was sorted by DE and CV.

The first step in selecting the best model parameters to simulate the spatial patterns of a study site was to define a "target" BC. Target BC was the estimated BC of the study site. To make this estimate, the expert system compared the proportion of water pixels on each border to the proportion of water pixels in the marsh as a whole. Those borders having a higher proportion of water pixels than the entire site were assumed to be influenced by a major water body at the border. The BC estimates were confirmed by visual examination of black-and-white photographs of binary land-water images of the sites. In a few cases, estimates were changed on the basis of the visual examination.

Once a target BC was selected, the solution file specific to the spatial pattern indices of that study site was searched for the "best" weighted mean W, specific to calculated G, for that BC. If a solution having the target BC was found in the group of solutions with the highest

DE, it was selected as the best solution. If more than one solution in the group of solutions with the highest DE had the target BC, then the one with the lowest CV was selected. If a solution having the target BC could not be found within the group having the highest DE, then the expert system sought a solution with the target BC among all solutions having DE within 75% of the highest DN. The solution having the target BC, the largest DE, and the lowest coefficient of variation was selected. If a solution having the target BC was not found in either of the above groups, then solutions having alternative BC were considered. First, solutions with BC having no more than one border different from that of the target BC were considered. Then, solutions having no more than two borders different from that of the target BC were considered. We usually found a solution having the target BC or no more than one border different from that of the target BC were considered ifferent from that of the target BC were considered.

References

- Adams, R. D., and R. H. Baumann. 1980. Land-building in coastal Louisiana: emergence of the Atchafalaya Bay delta. Center for Wetland Resources, Louisiana State University, Baton Rouge.
- Baumann, R. H. 1980. Mechanisms of maintaining marsh elevaion in a subsiding environment. Master's thesis, Louisiana State University, Baton Rouge.
- Browder, J. A., H. A. Bartley, and K. S. Davis. 1984. A probablilistic model of hte relationship between marshland-water interface and marsh disintegration. Ecological Modelling 29:245-260.
- Butera, M. K. 1982a. Identification of water bodies using remotely sensed multispectral scanner data, with applications for inventory, hydrologic assessment, and habitat evaluation. NASA Tech. Memo. No. TM-84672, NASA Earth Resources Laboratory, NSTL, Miss.
- Butera, M. K. 1982b. A distance measurement derived from Landsat MSS data, with application to marsh productivity, efficient crop transport, and environmental distrubance assessment. NASA Tech. Memo. No. TM-84671, NASA Earth Resources Laboratory, NSTL, Miss.
- Butera, M. K., and B. R. Seyfarth. 1981. A determination of marsh detritial export from Landsat-MSS data--a function of transport distance and water-body characterization. Proc. Machine Processing of Remotely Sensed Data Symposium. Purdue University, West Lafayette, Indiana.
- Butera, M. K., A. L. Frick, and J. A. Browder. 1984. A preliminary report on the assessment of wetland productive capacity from a remote-sensing-based model--a NASA/NMFS joint research project. IEEE Transactions in International Geoscience and REmote Sensing GE-22:502-511.
- Chabreck, R. H. 1972. Vegetation, water and soil characteristics of the Louisiana coastal region. Agricultural Experiment Station Bull. 664. Louisiana State University, Baton Rouge.

Cliff, A. D., and J. K. Ord. 1973. Spatial Autocorrelation. Pion, London, England.

Cowardin, L. M., V. Carter, F. C. Golet, and E. T. LaRoe. 1979. Classification of Wetlands and Deepwater Habitats of the United States. FWS/OBS-79/31, Office of Biological Services, U.S. Fish and Wildlife Service, Washington, D.C.

- Dow, D. D. 1982. Software programs to measure interface complexity with remote-sensing data, with an example of a marine ecosystem application. NASA Report No. 219. NASA Earth Resources Laboratory, NSTL, Miss.
- Dow, D. D., and R. W. Pearson. 1982. SLIN: a software program to measure interface length. NASA Report No. 208. NASA Earth Resources Laboratory, NSTL, Miss.
- Dozier, M. D. 1983. Assessment of change in the marshes of southwestern Barataria Basin, Louisiana, using historical aerial photographs and a spatial information system. Master's thesis, Louisiana State University, Baton Rouge.
- Faller, K. H. 1977. A procedure for dteection and measurement of interfaces in remotelyacquired data using a digital computer. NASA Tech. Report No. TR R-472. Lyndon B. Johnson Space Center, Houston, Tex.
- Faller, K. H. 1979. Shoreline as a controlling factor in commercial shrimp production. NASA Tech. Memo. 72-732. NASA Earth Resources Laboratory, NSTL, Miss.
- Gagliano, S. M., K. J. Meyer-Arendt, and K. M. Wicker. 1981. Land loss in the Mississippi River detlaic plain. Transactions of the Gulf Coat Association of Geological Societies 31:295-300.
- Gosselink, J. G. 1984. The ecology of the delta marshes of coastal Louisiana: a community profile. FWS/OBS-84/09. Office of Biological Services, U.S. Fish and Wildlife Service, Washington, D.C.
- Gosselink, J. G., C. C. Cordes, and J. W. Parsons. 1979. An ecological characterization study of the Chenier Plain coastal ecosystem of Louisiana and Texas. FWS/OBS-78/9 through 78/11 (3 vols.). Office of Biological Services, U.S. Fish and Wildlife Service.
- Graham, M. H., B. G. Junkin, M. T. Kalcic, R. W. Pearson, and B. R. Seyfarth. 1984. ELAS: Earth Resources Laboratory Applications Software. Vol. 2, ELAS User's Guide. NASA Earth Resources Laboratory, NSTL, Miss.
- Hicks, S. D. 1981. Long-period sea-level variations in the United States through 1978. Shore and Beach 49:26-29.
- Hutchnison, G. E. 1957. A Treatise in Limnology. Vol. 1, part 1. John Wiley, New York.
- Johnson, W. B., and J. G. Gosselink. 1982. Wetland loss directly associated with canal dredging in the Louisiana coastal zone. Pp. 60-72 in D. F. Boesch, ed., Proceedings of the Conference on Coatal Erosion and Wetland Modification in Louisiana: Causes,

Consequences, and Options. FWS/OBS-82/59. Office of Biological Services, U.S. Fish and Wildlife Service, Washington, D.C.

- Leibowitz, S. G., and J. M. Hill. 1988. Spatial analysis of Louisiana coastal land loss. In R. E. Turner and D. R. Cahoon, eds., Causes of Wetland Loss in the Coastal Central Gulf of Mexico. OCS Study/MMS 87-0119. Minerals Management Service, New Orleans.
- Mandelbrot, B. B. 1967. How long is the coast of Britain? Statistical self similarity and fractional dimension. Science 56:636-638.
- Moran, P. A. P. 1948. The interpretation of statistical maps. Journal of the Royal Statistical Society, Series B, 10:243-251.
- Pun, T. 1981. Entropic thresholding, a new approach. Computer Graphics and Image Processing 16:210-239.
- Richardson, L. F. 1961. The problem of contiguity: an appendix of statistics of deadly quarrels. General Systems Yearbook 6:139-187.
- Rosen, P. S. 1980. Erosion susceptibility of the Virginia Chesapeake Bay shoreline. Marine Geology 34:45-59.
- Sasser, C. E. 1977. Distribution of vegetation in Louisian coastl marshes as response to tidal flooding. Master's thesis, Louisiana State University, Baton Rouge.
- Sasser, C. E., M. D. Dozier, J. G. Gosselink, and J. M. Hill. 1986. Spatial and temporatl changes in Louisiana's Barataria Basin marshes, 1945-1980. Environmental Management 10:671-680.
- Turner, R. E., K. L. McKee, W. B. Sikora, J. P. Sikora, I. A. Mendelssohn, E. Swenson, C. Neill, S. G. Leibowitz, and F. Pedrazini. 1984. The impact and mitigation of man-made canals in coastal Louisiana. Nat. Sci. Tech. 6:497-504.
- Tuttle, J. R., and A. J. Combe. 1981. Flow regime and sediment load affected by alterations of the Mississippi River. Pp. 334-348 in R. D. Cross and D. L. Williams, eds., Proc. National Symposium on Freshwater Inflow to stuaries. FWS/OBS-81/04. Office of Biological Services, U.S. Fish and Wildlife Service, Slidell, La.
- Upton, G., and B. Fingleton. 1985. Spatial Data Analysis by Example. Vol. 1. John Wiley, New York.

- Wicker, K. M. 1980. Mississippi deltaic plain region ecological characterization: a habitat mapping study. A User's Guide to the Habitat Maps. FWS/OBS-79/07. Office of Biological Services, U.S. Fish and Wildlife Service, Slidell, La.
- Zimmerman, R. J., T. J. Minello, and G. Zamora, Jr. 1984. Selection of vegetated habitat by brown shrimp, *Penaeus aztecus*, in a Galveston Bay salt marsh. U.S. Fisheries Bull. 82:325-336.

SLIN output	Class code	SLIN output	Class code	SLIN output	Class code
0	5	24	2	48	3
1	0	25	1	49	3 3 2 2 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 4
2	0	26	1	50	2
3	1	27	2	51	2
4	0	28	2	52	2
5	0	29	2	53	3
6	1	30	2	54	3
7	1	31	2	55	3
1 2 3 4 5 6 7 8 9	0	32	2	56	3
	1	33	2	57	3
10	1	34	2	58	3
11	0	35	1	59	3
12	2	36	2	60	3
13	1	37	2	61	3
14	1	38	2	62	3
15	1	39	2	63	3
16	1	40	2	64	3
17	1	41	2	65	4
18	1	42	2	66	4
19	2 2	43	2	67	4
20		44	2	68	4
21	1	45	2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	69	4
22	1	46	3	70	ND
23	2	47	3	71	ND

Table A1.Look-up table used to classify water and land identified by the ELAS
shoreline length (SLIN) module into water pixels and land pixels with zero,
one, two, three, and four sides adjacent to water.

Key to Class Codes: 0 = 1 and pixel with zero sides adjacent to water.

1 = land pixel with one sides adjacent to water.

2 = land pixel with two sides adjacent to water.

3 = land pixel with three sides adjacent to water.

4 = land pixel with four sides adjacent to water.

5 = water pixel.

Ind	ex			
Name	Weight	W	W • Weight	
Interface	4,011	60	240,660	
Adj-0	108	53	5,724	
Adj-1	805	256	206,080	
Adj-2	1,257	121	152,097	
Adj-3	1,109	53	58,777	
Adj-4	680	156	106,080	
-	7,970		769,418	
Weighted mea Decision Num CV = 64.91%	n W = 97 ber (DE) = 7,9	970		
(Number of b (Σ Adj weigh	order pixels = ts = 3,959)	= 52)		

Table A2. Calculation of weighted mean W, DE, and CV from a G-BC look-up table in which G = 0, BC = 0000, and DL = 10.90%.

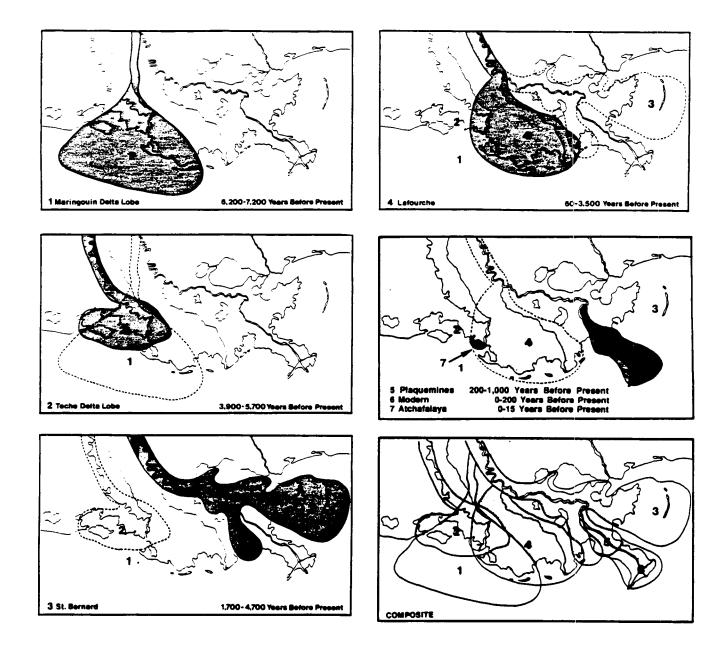


Figure A1. The maximum extent of the influence of deltaic lobes of the Mississippi River on the present geomorphology of Louisiana's coastal wetlands (modified from Adams and Baumann 1980).

Appendix **B**

Spatial-Pattern Statistics of Simulations Using All Combinations of W, G, and BC

- Table B1. Interface length (number of pixels) vs. W, G, and border condition (BC) at 50% disintegration level (DL), from simulations of 192-x-192-pixel marshes.
- Table B2. Disintegration level (DL) at maximum interface for simulations of 192-x-192 pixel marshes with varying W, G, and border conditions.

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				Border	Condition		
W	G	0000	0001	0011	0101	0111	1111
0 4 20 60 180 540	0 0 0 0 0	36,695 24,946 18,243 14,357 10,921 8,113					
0 4 20 60 180 540	4 4 4 4 4		34,573 24,801 18,516 14,434 10,726 7,672	33,427 24,630 18,115 14,300 10,959 7,957	33,412 24,505 18,392 14,595 10,443 8,183	31,888 24,243 18,264 14,385 11,053 8,156	30,754 24,300 18,296 14,384 11,272 8,254
0	20		31,208	26,080	28,635	20,395	16,088
4	20		23,401	21,440	22,500	19,311	16,768
20	20		17,924	17,271	17,444	16,867	16,612
60	20		14,273	13,863	14,167	13,766	13,747
180	20		10,937	11,055	10,719	10,999	10,586
540	20		7,940	8,039	7,706	8,058	8,428
0	60		25,752	15,520	20,946	9,667	6,998
4	60		20,540	15,743	18,578	11,001	8,120
20	60		16,791	14,475	16,031	12,348	10,069
60	60		13,777	13,008	13,366	11,792	10,904
180	60		10,771	10,227	10,343	9,927	9,561
540	60		7,789	7,679	7,732	7,895	8,075
0	180		16,113	6,001	9,302	3,836	2,946
4	180		15,994	6,850	10,708	4,251	3,282
20	180		13,866	9,012	11,416	5,402	3,914
60	180		12,481	9,622	11,457	6,687	5,722
180	180		9,919	8,821	9,644	7,551	6,054
540	180		7,789	7,556	7,524	7,404	6,622
0	540		5,961	2,468	3,831	1,939	1,846
4	540		7,093	2,534	4,121	1,979	1,922
20	540		8,826	3,168	5,530	2,205	1,946
60	540		9,322	4,650	7,169	2,737	2,220
180	540		8,513	5,826	7,248	3,693	2,772
540	540		6,877	5,408	6,890	4,572	3,900

Table B1. Interface length (number of pixels) vs. W, G, and border condition (BC) at 50% disintegration level (DL), from simulations of 192-x-192-pixel marshes.

0000 = no water borders, 0001 = one water border, 0011 = two adjacent water borders, 0101 = two opposite water borders, 0111 = three water borders, 1111 = four water borders.

				Border C	ondition		
W	G	0000	0001	0011	0101	0111	1111
0 4 20 60 180 540	0 0 0 0 0	49.56 47.24 49.99 45.10 52.40 52.55					
0 4 20 60 180 540	4 4 4 4 4		46.51 47.42 49.90 49.66 47.59 57.06	45.35 48.99 50.24 49.53 50.33 44.78	44.74 48.76 47.99 50.53 46.68	43.98 47.17 47.35 51.21 57.19 52.37	42.64 47.71 50.55 48.13 55.88 54.76
0 4 20 60 180 540	20 20 20 20 20 20		45.11 47.78 52.63 47.39 49.70 48.21	49.76 49.98 50.43 52.37 49.25 45.57	46.42 49.18 51.19 51.41 54.64 42.41	52.42 51.18 47.46 48.01 51.47 49.38	48.42 50.76 48.57 48.37 51.36 49.13
0 4 20 60 180 540	60 60 60 60 60		49.02 47.17 50.50 52.80 50.49 47.00	57.92 54.83 47.74 50.51 50.81 49.67	50.40 51.14 52.20 54.88 50.47 54.79	49.56 51.94 54.46 49.13 49.37 46.17	40.26 51.91 49.21 53.12 44.85 50.40
0 4 20 60 180 540	180 180 180 180 180 180		57.35 55.18 48.04 49.80 51.01 51.84	59.07 57.84 55.26 49.30 47.94 52.39	57.26 57.12 56.49 51.49 54.02 56.04	49.01 53.10 59.55 55.46 50.60 48.08	36.41 33.12 53.95 56.67 51.47 44.68
0 4 20 60 180 540	540 540 540 540 540 540		56.58 59.41 59.55 50.65 52.96 54.74	49.25 55.14 59.64 62.83 62.62 47.00	55.69 61.17 50.22 49.95 57.95	31.70 39.76 44.34 68.04 61.40 56.73	11.22 13.02 17.11 15.17 47.89 53.37

Table B2. Disintegration level (DL) at maximum interface for simulations of 192-x-192 pixel marshes with varying W, G, and border conditions.

0000 = no water borders, 0001 = one water border, 0011 = two adjacent water borders, 0101 = two opposite water borders, 0111 = three water borders, 1111 = four water borders.

Appendix C

Spatial-Pattern Statistics of TM Scenes

Table C1. Values of the spatial-pattern indices measured in the TM imagery.

Table C-1. Values of the spatial pattern indices measured in the TM imagery: interface length, the side-adjacency statistics (number of water pixels with 0, 1, 2, 3, and 4 sides adjacent to other water pixels), and number of water pixels on the marsh border.

	– Quađ	Quarter	ter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
	_ Late Lafo	<u>Lafourche</u> ,	, salt								
	Leeville		MN	42.94	,09		e e	, 12	,96	, 85	<u> </u>
			NE	34.93	,02		Ч	, 31	,47	, 86	വ
			SE	39.91	10,118	146	519	2,470	2,901	8,369	308
			MS	46.74	, 11		ß	,49	, 06	,06	0
	Mink Bayou	Ŋ	MN	26.59	, 16	0	Ο	, 68	, 06	, 32	Ч
đ			NE		, 51	Ч	4	, 31	, 28	,74	ŝ
51			SE		,42		Ч	,87	, 97	, 97	ŝ
			SW	29.57	, 93	0	8	,86	, 19	, 02	2
	Caminada Pass	Pass	MN		, 77		4	, 66	, 99	8,30	σ
			NE		, 34		2	Г	н	9,05	4
			SE	9.2	10			e	2	5,79	4
			SW	58.02	, 88		254	D	0	, 84	1
	Bay Tambour	our	MN	1.1	Ч	61	e	0	, 78	4,89	ω
			NE	76.55	, 04		92	2	Ч	5,71	0
			SE	8.7	, 24		73	e	2	0,54	0
			SW	84.38	, 51		106	e	9	8,12	S
	Pelican F	Pass	MN	<u>م</u>	4	н	11	9	9	5,49	ഹ
			NE	82.60	c	7	93	7	e	8,22	Ч
			SE		, 10	13	64	0	c	9,91	2
			SW	97.24	ω	0	4	0	9	4,83	Ω

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Table

– Quad Qua	Quarter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
<u>Early Lafourche, salt</u>	<u>che, sa</u>	<u>1t</u>							
Grand Bayou	MN	53.83	,43	48	7	4	, 15	7,37	
du Large	NE	67.26	3,621	37	158	886	1,145	22,053	514
	SE	40.44	,17	109	7	ω	, 83	0,34	4
	SW	83.72	, 11	17	50	S	9	9,64	c
Lake La	MN	91.13	, 07	17	54	2	S	1,51	7
Graisse	NE	99.28	σ	0	e	67	4	5,64	4
Central Isles		75.77	e	31		0	9	5,01	7
Dernieres	NE	90.05	, 60	9		e	0	1,41	0
	SE	95.18	4	10		ω	S	3,91	σ
	MS	93.53	4	ñ		7	e	3,24	Ч
Cocodrie	MN	33.43	,43	106	σ	, 69	, 00	7,76	9
	NE	68.89	ß	40		S	σ	, 59	7
	SE	85.79	, 68	32	ß	Ч	, 38	8,65	ω
	MS	54.72	, 69	69	Ч	, 25	, 51	3,52	0
Dog Lake	MN	62.97	, 61	29	ß	ω	,76	9,78	8
	NE	30.74	, 11	113	4	, 89	, 39	,21	9
	SE	42.38	77,	95	0	, 75	, 53	, 51	H
	MS	36.40	,48	67	e	, 56	,96	,10	æ

Table C-1. (cont. 3)

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- Quad Quarter	ter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
1									
<u>Late Lafourche, brackish</u>	, brac	<u>kish</u>							
Lake Bully	MN	33.69	-	46	2	, 25	,07	, 48	5
Camp	NE	47.84	•	0	9	, 03	, 84	, 74	4
I	SE	22.21	•	122	-	, 36	, 57	, 48	e
	SW	50.02		78	Q	, 12	, 37	, 98	0
Golden Meadow	MN	30.91	•	81	-	, 53	, 17	, 93	9
Farms	NE	60.69	5,700	80	367	1,181	1,836	18,357	553
	SE	•	•	86	-	, 81	, 53	7,86	σ
	MS	38.95	•	47	4	, 08	<i>LL</i>	1,04	വ
Bay L'Ours	SE	81.53	•	42	വ	2	963	, 35	Г
	MS	52.26		34	0	ω	513	7,83	ω
Three Bayou	MN	32.93	•	108	9	2	,10	7,41	2
Вау	NE	58.71	•	60	Ч	7	1,218	77,	σ
Golden Meadow	MS	27.66	•	182	0	Ы	, 18	, 80	0

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	(conte -	1.							
- Quad Quarter	ter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Ađj-4	Border Pixels
<u>Early Lafourche, brackish</u>	le, bra	<u>ackish</u>							
Lost Lake	MN	32.65	, 09	62	Ň	, 35	, 07	, 03	
	NE	٠	,20	107	S	, 06	, 84	, 57	σ
	SE	21.81	7,227	2	496	1,644	1,820	3,689	265
	MS	•	,64	75	4	,91	, 07	2,87	4
Lake Mechant	MN	•	,07	69	2	, 44	, 83	,40	e
	NE	•	,78	06	5	, 61	, 95	6,71	9
	SE	•	, 57	37	9	, 19	,47	, 30	α
	MS	٠	,05	23	2	82	89	2,31	വ
Bayou Sauveur	MN	8.72	, 57	82	σ	77	79	, 23	4
	NE	٠	, 84	160	S	, 20	, 03	96	0
	SE	٠	, 58	79	2	, 30	, 55	, 72	0
	MS	•	, 89	36	σ	, 24	, 59	,74	4
Lake Quitman	NE	٠	,98	69	4	60	2	10	£
	SE	٠	,98	64	0	,42	, 88	, 03	4
	ΜS	•	, 54	63	9	, 57	,94	,49	0
Dulac	NE	٠	,70	82	σ	σ	88	,46	2
	SE	•	,71	30	Ч	Q	85	, 73	2
Montegut	SE	•	, 31	112	2	2	ഹ	2,07	349
	SW	27.16	, 78	67	2	, 52	, 10	, 66	e

Table C-1. (cont. 4)

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Appendix D

Simulation Results

Table D1. W, G, and BC selected by the expert system to simulate each TM scene.

- Table D2. Values of spatial-pattern indices from 70 simulations.
- Table D3. Simulation values for interface length and disintegration level compared with TM scene disintegration levels.
- Table D4. Coefficients of variation from three sets of three replicate simulations using the same values of W, G, and BC.
- Figure D1. "Best-fit" simulations showing interface as a function of disintegration level for 70 marsh sites in coastal Louisiana.
- Figure D2. Coefficients of variation of spatial-pattern indices from three sets of three repeated simulations with the same W, G, BC, and DL values.

	leve	el (]	DL), tar	get BC,	decis	sion nu	umber ((DE), and pixels [TW	CV.
							Target		
Quad	Quar	ter	DL	W	G	BC	BC	DE/TWP	CV
Late Laf	fourche	e, sa	<u>lt</u>				<u></u>	· · · ·	
Leeville	9	NW	42.94	272	60	0011		195.1	42
		NE	34.93	188	0	0111		197.2	43
		SE	39.91	237	4	0011		194.4	42
		SW	46.74	8	180	0011		196.8	128
Mink Bay	ou	NW NE	26.59	244	180	0101		197.8	46
		SE	24.54 29.44	24 248	180 0	0011 0111		197.4 197.8	115
		SW	29.57	240	180	0001		197.8	58 45
Caminada	Pass	NW	62.09	58	180	0111		190.6	4J 9
		NE	82.87	13	540	0111	0101	197.8	78
		SE	99.27	1	540	0011		198.0	225
		SW	58.02	28	540	0011		197.6	32
Bay Tamb	our	NW	51.18	31	180	0111		190.0	41
		NE	76.55	17	540	0011		198.2	84
		SE SW	88.72 84.38	107 13	540 180	0111		197.9	53
Pelican	Pass	NW	98.96	8	180	0111 0111		197.6 197.9	74 56
	- 455	NE	82.60	88	540	0111		198.0	35
		SE	86.68	16	540	0011		197.8	76
		SW	97.24	38	540	1111	0111	197.9	49
<u>Early La</u>	fourch	<u>e, s</u>	<u>alt</u>						
Grand Ba	you	NW	53.83	30	540	0011	1	198.4	40
du Lar	ge	NE	67.26	29	540	0011		197.9	36
		SE	40.44	14	180	0011		197.0	95
- 1 -		SW	83.72	0	540	0011	0001	98.1	211
Lake La	-	NW	91.13	113	540	0111		198.0	55
Graiss Central		NE	99.28	10	180	0111		197.8	71
Dernie		NW NE	75.77 90.05	26 43	180 540	1111		197.1	26
Dermie	169	SE	95.18	43 16	540 540	0111 0111	0101	197.8 198.0	70 26
		SW	93.53	33	540	1111	0101	197.9	4 3
Cocodrie	:	NW	33.43	115	180	0101	••••	193.9	72
		NE	68.89	311	540	0001		197.2	55
		SE	85.79	69	540	0011		195.6	15
		SW	54.72	233	60	0011		196.0	30
Dog Lake		NW	62.97	44	540	0011	0001	198.3	65
		NE SE	30.74	213	180	0101		193.7	32
		SE SW	42.38 36.40	120 123	180 540	0011	0101	198.0	56
		5	JU.40	143	540	0001	0101	197.1	61

Table D1. W, G, and border condition (BC) selected by the expert system to simulate each TM scene, with disintegration level (DL), target BC, decision number (DE), and CV. (DE is expressed as percentage total water pixels [TWP].)

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Quad Q	uarter	DL	W	G	T BC	arget BC	DE/TWP	CV
Late Lafour	che, br	ackish						
Lake Bully	NW	33.69	62	180	0111		197.8	90
Camp	NE	47.84	244	180	0001		198.0	52
-	SE	22.21	130	60	0011		192.1	88
	SW	50.02	255	60	1111		197.8	41
Golden Mead	low NW	30.91	129	180	0011		197.7	45
Farms	NE	60.69	160	540	0011		197.5	14
	SE	35.56	308	60	0011		193.8	47
• -	SW	38.95	3,184	0	0000	0001	198.9	53
Bay L'Ours	SE	81.53	8	180	0111		197.9	58
	SW	52.26	1	540	0011		98.0	807
Three Bayou		32.93	100	60	1111	0001	196.9	80
Bay	NE	58.71	33	540	0011 0011	0001	198.2	43 82
Golden Mead	low SW	27.66	116	0	0011		198.0	82
Early Lafou	irche, b	rackis	h					
Lost Lake	NW	32.65	23	180	0011		198.4	131
	NE	47.32	245	180	0001		198.3	53
	SE	21.81	111	60	0111		196.7	66
	SW	50.78	290	20	0111		195.8	61
Lake Mechan	t NW	64.05	4	180	0011		196.4	132
	NE	29.91	118	60	0011		194.1	86
	SE	44.91	86	540	0011		190.5	35
	SW	67.06	17	540	0011		197.8	81
Bayou Sauve		8.72	325	0	0000		198.8	78
	NE	10.90	93	0	0000		197.4	61
	SE	22.23	133	180	0011		197.5	45
	SW	24.31	35	540	0001		198.4	141
Lake Quitma		48.20	116	540	0001		196.3	75
	SE	32.41	20	180	0011		197.9	110
						0111		52
Dulac							199.2	39
Montegut								59
	SW	27.16	404	U	0000		197.6	70
Dulac Montegut	SW NE SE SW	31.85 9.63 66.80 47.48 27.16	121 701 10,947 289 404	180 0 180 0	0011 0000 0000 0101 0000	0111	197.4 199.2 195.3 197.6	

Table D-1. (cont.)

Key to BC: 0 = land border, 1 = water border. For example, 0101 indicates two opposite water borders. The maximum possible value of 100 DE/TWP is slightly less than 200. Table D-2. Values of spatial pattern indices from the 70 simulations: interface length, side-adjacency statistics (number of water pixels with 0, 1, 2, 3, and 4 sides adjacent to other water pixels), and number of water pixels on the marsh border. Disintegration level (DL, water area as percentage total area) is also shown.

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Quad Qua	Quarter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Late Lafourche,	ne, salt	1							
Leeville	MN	42.94	, Ч	21	୦	, 85	2	, 55	443
	NE	34.93	,93	17	Ο	, 04	2	, 53	σ
	SE	39.91	,81	16	9	,94	, 44	, 29	ഹ
	SW	46.74	7,780	ß	1,174	1,026	721	13,460	497
Mink Bayou	MN	26.59	,24	658	വ	,45	Ч	00,	ŝ
	NE	24.54	, 35	c	0	9	73	, 77	N
	SE	29.44	, 68	18	e	,74	, 84	, 39	Ч
	SW	29.57	,16	18	σ	63	7	, 58	ω
Caminada Pass		62.09	,98	33	9	, 39	,04	8,07	7
	NE	82.87	, 32	76	e	9	ω	9,18	0
	SE	99.27	19	12	Ч	2	വ	5,76	ŝ
	SW	58.02	,94	Ο	e	ŝ	ω	9,00	0
Bay Tambour	MN	51.18	ω	111	4		S	5,20	4
	NE	76.55	, 05	74	З	7	2	6,19	7
	SE	88.72	, 25	7	9	S	S	0,54	Ч
	SW	84.38	<i>TL</i>'	70	4	9	σ	8,41	0
Pelican Pass	MN	98.96	e	1	e	α	9	, 35	4
	NE	82.60	σ	11			9	8,28	0
	SE	86.68	, 35		0	Ч	Ч	0,16	Ч
	SW	97.24	4	7			c	4,82	9

Quad Qua	Quarter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Early Lafourche, salt	che, sa]	Ľ.							
Grand Bayou	MN		σ	1	- 	ેન	4	7,55	
du Large	NE	67.26	4	7	Ċ	0	2	2,21	4
	SE	40,44	7,978	298	1,077	1,277	937	10,845	471
	MS	83,72	7	9	ω	Ô	S	9,64	Ч
Lake La	MN	91,13	, 03		148		4	1,56	Г
S		9+2	σ		2	Ω.	0	5,45	S
Central Isles		5+7	4		382	Ó	e	5,22	9
Dernieres	NE	0,0	Ч		4	N	9	1,72	2
	SE	5 + 1	1		5	2	Ч	3,93	e
	MS	3,5	4		2	9	2	3,23	9
Cocodrie	MN	3,4	7		4	,94	,74	6,28	9
	NE	8,8	, 11		e	σ	e	0,19	e
	SE	5;7	, 24		N	ŝ	, 16	8,96	σ
	ΜS	4,7	,27		δ	N	, 75	3,49	σ
Dog Lake	MN	2.9	, 39		ω	Ó	95	0,32	e
	NE	5	N	25	710	, 66	, 54	, 93	S
	SE	42,38	,43		σ	e	2	, 38	ß
	SW		,61		0	,71	, 54	,99	2

Table D-2. (cont. 2)

Table D-2. (cont. 3)

Quad Quarter	ter	DĽ	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Late Lafourche, brackish	, brac	kish							
Lake Bully	MN	33.69	8	71	~	, 12	, 24	, 68	- -
Camp	NE	47.84	°,	6	9	, 80	,40	, 40	S
	SE	22.21	8,350	44	805	1,732	2,224	2,951	431
	SW	•	e,		7	, 84	, 61	, 53	S
Golden Meadow	MN	٠	°,	41	9	, 69	, 18	6,28	ŝ
Farms	NE	•	2		Ч	, 25	, 07	, 99	ч
	SE	35.56	ц С	19	ω	, 75	, 87	7,32	9
	SW		4,	e	Ч	S	, 73	1,36	σ
Bay L'Ours	SE	81.53	e,	107	4	Ч	0	, 79	δ
	SW	52.26	4	ω	c	2	S	7,85	δ
Three Bayou	SE	32.93	č,	34	e	8	8	5,74	ഹ
Bay	MS	58.71	4,108	85	e	4	Ś	, 11	0
Golden Meadow	SW	27.66	10,187	45	0	7	3,029	4,00	129

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(cont.
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Table

Quad Quarter	ter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Early Lafourche, brackish	e, bra	<u>ckish</u>							
Lost Lake	MN	32.65	57		1,088	, 22	, 00	, 05	ŝ
	NE	47.32	, 38		67	, 86	, 51	, 03	4
	SE	21.81	, 24		821	2	15	2,75	ω
	MS	50.78	S		620	, 85	, 54	2,13	
Lake Mechant	MN	64.05	69		1,137	2	σ	,94	ഹ
	NE	٠	, 79		857	σ	, 00	4,66	9
	SE	٠	, 84		526	4	α	3,46	9
	SW	67.06	, 22		440	7	σ	, 75	2
Bayou Sauveur	MN	8.72	, 99		430	2	4	ഹ	4
	NE	10.90	, 31		776	, 24	, 13	Ч	2
	SE	22.23	, 73		678	, 33	Э	, 88	N
	MS	24.31	, 31		903	, 06	2	, 73	α
Lake Quitman	NE	2	,46		773	ω	0	,47	
	SE	32.41	, 26		989	, 16	89	, 18	ß
	SW	31.85	,00		797	, 57	, 28	, 59	4
Dulac	NE	9.63	,44		311	0	, 02	,46	'n
	SE	66.80	, 85		182	Ο	, 26	2,31	9
Montegut	SE	47.48	8,551	19	670	1,674	3,087	11,578	473
	SW	27.16	,43		578	, 57	,41	, 30	2

Quad Qu	arter	Maximum Sim. Interface Length	DL at Sim. Maximum Interface	1985 TM DL	Sign of Diff. (SimTM)
Late Lafourc	<u>he, salt</u>				
Leeville	NW	9,329	51.8	42.0	+
	NE	10,681	47.6	34.0	+
	SE	10,156	49.3	39.0	+
	SW	8,638	63.8	46.0	+
Mink Bayou	NW	8,905	51.1	26.0	+
	NE	10,200	55.6	24.0	+
	SE	9,923	49.5	29.0	+
	SW	9,543	56.6	29.0	+
Caminada Pas		7,087	60.4	62.0	_
	NE	2,188	33.2	82.0	-
	SE	2,582	42.6	99.0	-
 .	SW	4,110	67.1	58.0	+
Bay Tambour	NW	6,232	65.7	51.0	+
	NE	3,444	60.2	76.0	
	SE	3,273	72.6	88.0	-
Deligen Dese	SW	5,086	59.6	84.0	-
Pelican Pass	NW	4,608	63.4	98.0	-
	NE	3,113	57.1	82.0	-
	SE Sw	3,265 2,417	59.0 10.9	86.0 97.0	-
Early Lafour	<u>che, salt</u>	2			
Grand Bayou	NW	4,010	60.1	53.0	+
du Large	NE	4,180	61.6	67.0	-
	SE	9,196	59.3	40.0	+
	SW	2,514	54.8	83.0	-
Lake La	NW	3,430	67.0	91.0	-
Graisse	NE	4,738	50.6	99.0	
Central Isle		4,442	53.1	75.0	-
Dernieres	NE	2,388	54.7	90.0	-
	SE	2,187	60.3	95.0	-
0 1 - '	SW	2,443	13.3	93.0	-
Cocodrie	NW	10,496	50.6	33.0	+
	NE	8,108	46.7	68.0	-
	SE	5,122	59.8	85.0	-
	SW	9,624	47.9	54.0	
Dog Lake	NW	4,473	59.4	62.0	-
	NE	9,375	53.4	30.0	+
	SE	9,662	45.5	42.0	+
	SW	9,737	51.8	36.0	+

Table D3. Simulation values for interface length and disintegration level (DL) compared with TM scene disintegration levels, with deviation and percentage deviation.

		Maximum Interface	DL at Maximum	1985	Sign
Quad Qua	rter	Length	Interface	DL	Diff.
	e breed	-i -h			
Late Lafourch	e, brack	<u>Kisn</u>			
Lake Bully	NW	7,055	61.7	33.0	+
Camp	NE	9,203	54.4	47.0	÷
	SE	10,787	57.4	22.0	+
	SW	9,410	49.6	50.0	
Golden Meadow		10,033	52.6	30.0	+
Farms	NE	6,393	57.8	60.0	
	SE	9,142	55.8	35.0	+
	SW	4,459	44.6	38.0	+
Bay L'Ours	SE	4,496	44.6	81.0	
_1 _	SW	2,582	42.6	52.0	-
Three Bayou	SE	9,835	46.6	32.0	+
Bay Golden Meadow	SW SW	4,218 12,284	56.8 50.3	58.0 27.0	 +
Early Lafourc	<u>he, brac</u>	<u>ckish</u>			
Lost Lake	NW	9,934	57.3	32.0	+
	NE	9,469	51.5	47.0	+
	SE	10,869	52.2	21.0	+
	SW	9,346	45.2	50.0	-
Lake Mechant	NW	7,717	63.2	64.0	-
	NE	11,116	53.2	29.0	+
	SE	5,717	59.0	44.0	+
			62.3	67.0	-
	SW	3,302		_	
Bayou Sauveur	NW	9,362	48.1	8.0	+
Bayou Sauveur	NW NE	9,362 13,189	48.1 48.7	10.0	+ +
Bayou Sauveur	NW NE SE	9,362 13,189 8,925	48.1 48.7 48.7	10.0 22.0	
-	NW NE SE SW	9,362 13,189 8,925 10,567	48.1 48.7 48.7 58.0	10.0 22.0 24.0	+
-	NW NE SE SW NE	9,362 13,189 8,925 10,567 9,602	48.1 48.7 48.7 58.0 48.7	10.0 22.0 24.0 48.0	+ + =
-	NW NE SE SW NE SE	9,362 13,189 8,925 10,567 9,602 9,235	48.1 48.7 48.7 58.0 48.7 55.3	10.0 22.0 24.0 48.0 32.0	+ + = +
Lake Quitman	NW NE SE SW NE SE SW	9,362 13,189 8,925 10,567 9,602 9,235 9,378	48.1 48.7 48.7 58.0 48.7 55.3 49.2	10.0 22.0 24.0 48.0 32.0 31.0	+ + = +
Lake Quitman	NW NE SE SW NE SE SW NE	9,362 13,189 8,925 10,567 9,602 9,235 9,378 7,600	48.1 48.7 58.0 48.7 55.3 49.2 49.5	10.0 22.0 24.0 48.0 32.0 31.0 9.0	+ + = +
Bayou Sauveur Lake Quitman Dulac	NW NE SW NE SW NE SE	9,362 13,189 8,925 10,567 9,602 9,235 9,378 7,600 3,425	48.1 48.7 58.0 48.7 55.3 49.2 49.5 41.2	10.0 22.0 24.0 48.0 32.0 31.0 9.0 66.0	+ + = +
Lake Quitman	NW NE SE SW NE SE SW NE	9,362 13,189 8,925 10,567 9,602 9,235 9,378 7,600	48.1 48.7 58.0 48.7 55.3 49.2 49.5	10.0 22.0 24.0 48.0 32.0 31.0 9.0	+ + = +

Table D-3. (cont.)

Spatial				Set		Moon
Pattern Index		1		2	3	Mean
Interfac	e	5.0	2	10.72	3.11	6.24
Adj-0		3.9	5	71.25	53.29	42.83
Adj-1		8.2	6	7.75		6.23
Adj-2		5.5	7	14.80	1.66	7.34
Adj-3		2.5	7	10.93	5.37	6.29
Adj-4		4.0	2	0.54	1.96	2.17
Mean		4.9	0	19.33	11.35	
Set 1:	W =	275,	G =	0,	BC = 1111,	DL = 29.44
Set 2:	w =	111,	G =	540,	BC = 1000,	DL = 88.72
Set 3:	= W	337,	G =	180,	BC = 1010,	DL = 47.48

Table D4. Coefficients of variation (CV) from three sets of three replicate simulations using the same values of W, G, and BC.

Figure D1. "Best-fit" simulations showing interface as a function of disintegration level for 70 marsh sites in coastal Louisiana. The ⊕ symbol indicates the coordinates of the comparable TM scene for each. Quadrangles are presented in the same order as in Table D1.

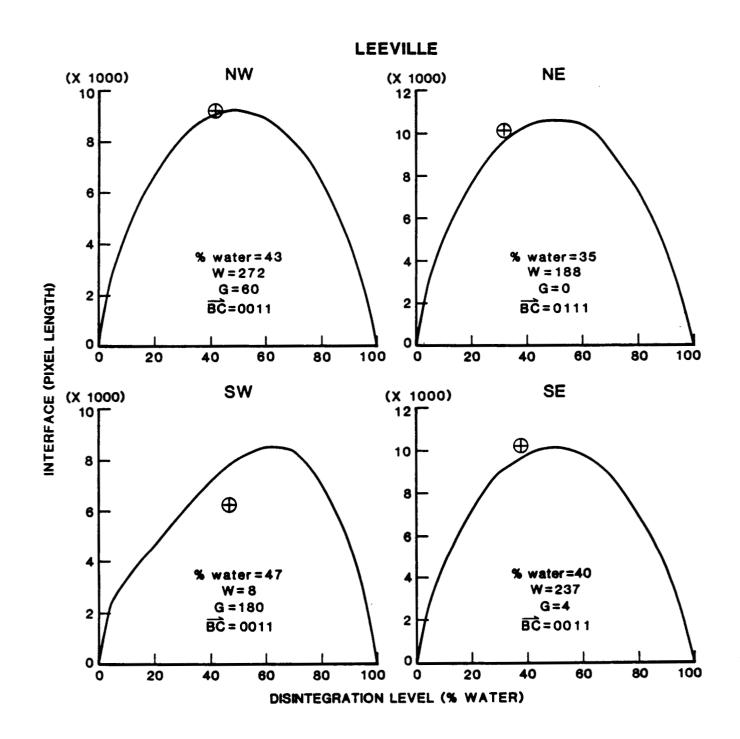


Figure D1-1. Simulated interface versus disintegration level, with TM scene coordinates, for the Leeville quadrangle.

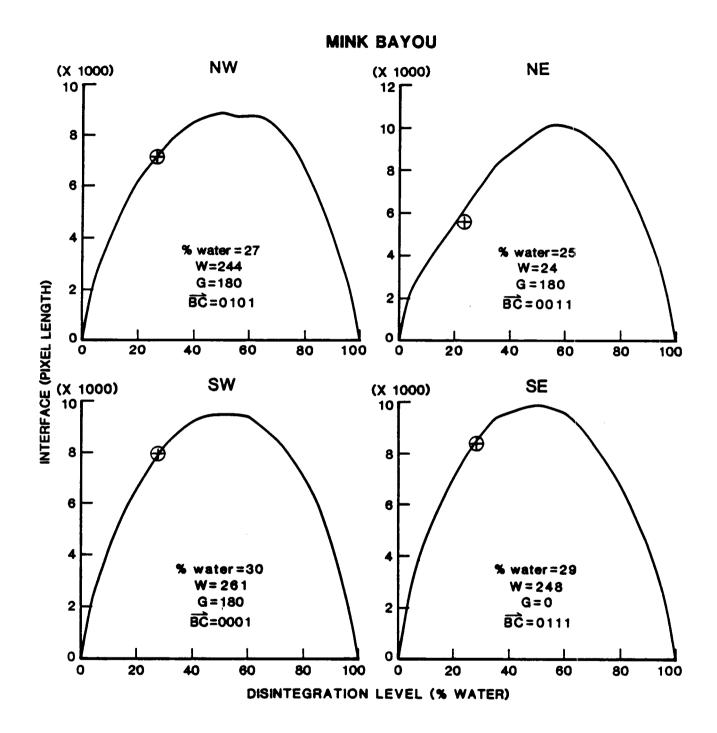


Figure D1-2. Simulated interface versus disintegration level, with TM scene coordinates, for the Mink Bayou quadrangle.

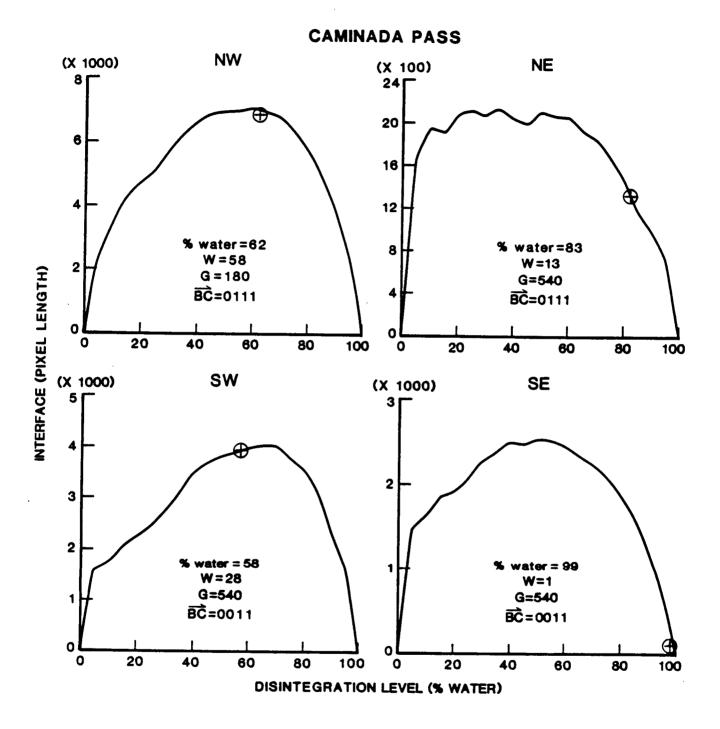


Figure D1-3. Simulated interface versus disintegration level, with TM scene coordinates, for the Caminada Pass quadrangle.

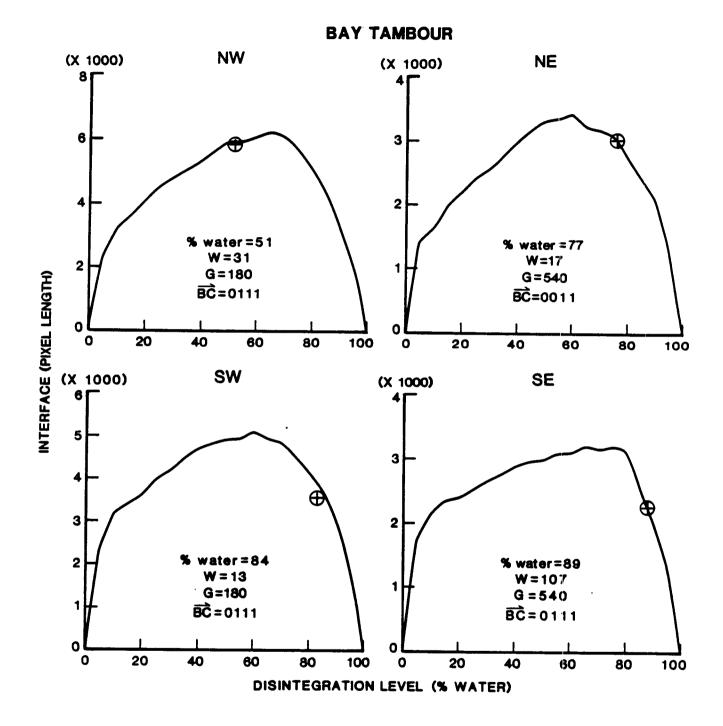


Figure D1-4. Simulated interface versus disintegration level, with TM scene coordinates, for the Bay Tambour quadrangle.

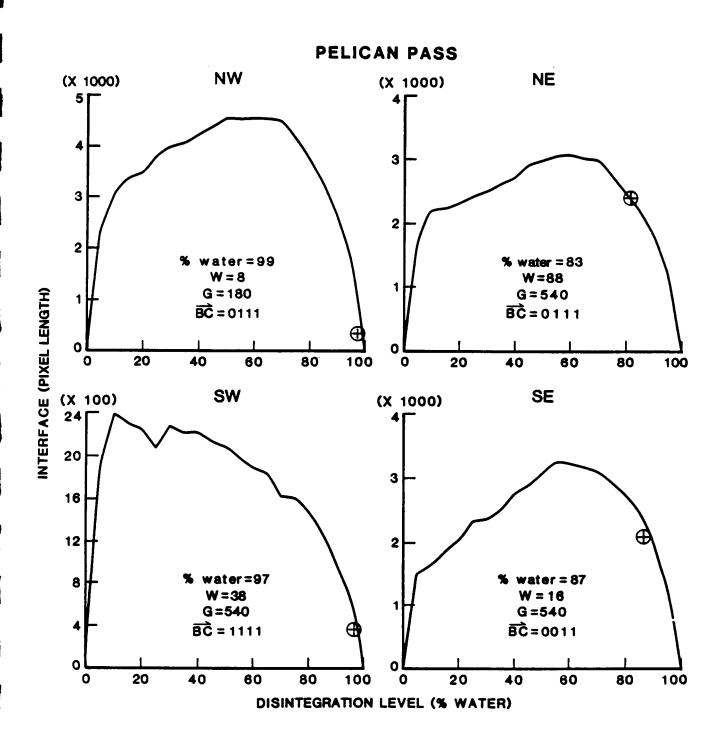


Figure D1-5. Simulated interface versus disintegration level, with TM scene coordinates, for the Pelican Pass quadrangle.

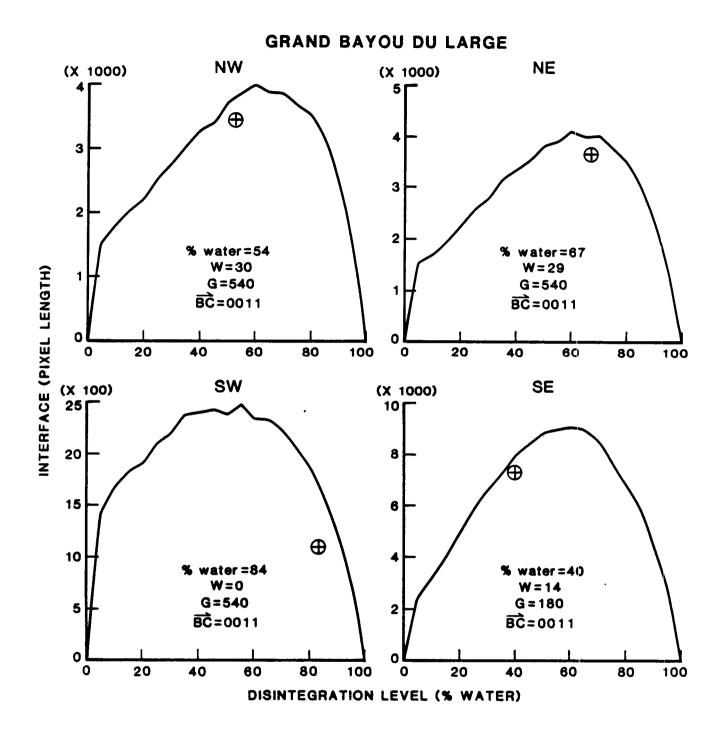


Figure D1-6. Simulated interface versus disintegration level, with TM scene coordinates, for the Grand Bayou du Large quadrangle.

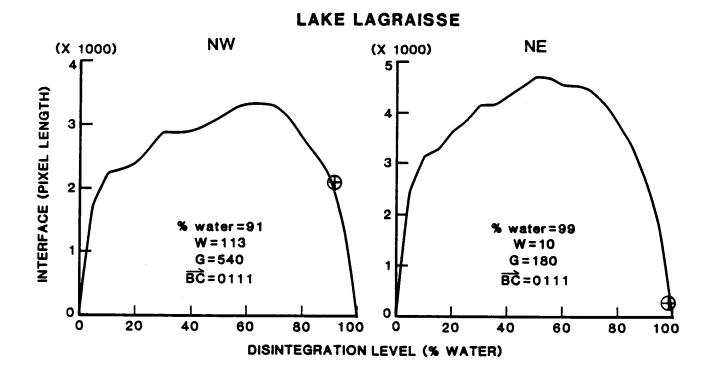


Figure D1-7. Simulated interface versus disintegration level, with TM scene coordinates, for the Lake La Graisse quadrangle.

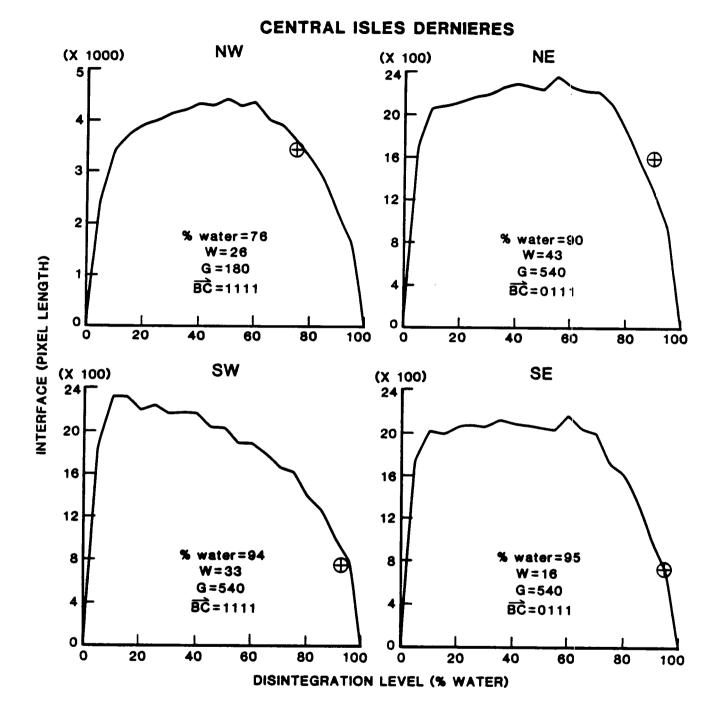


Figure D1-8. Simulated interface versus disintegration level, with TM scene coordinates, for the Central Isles Dernieres quadrangle.

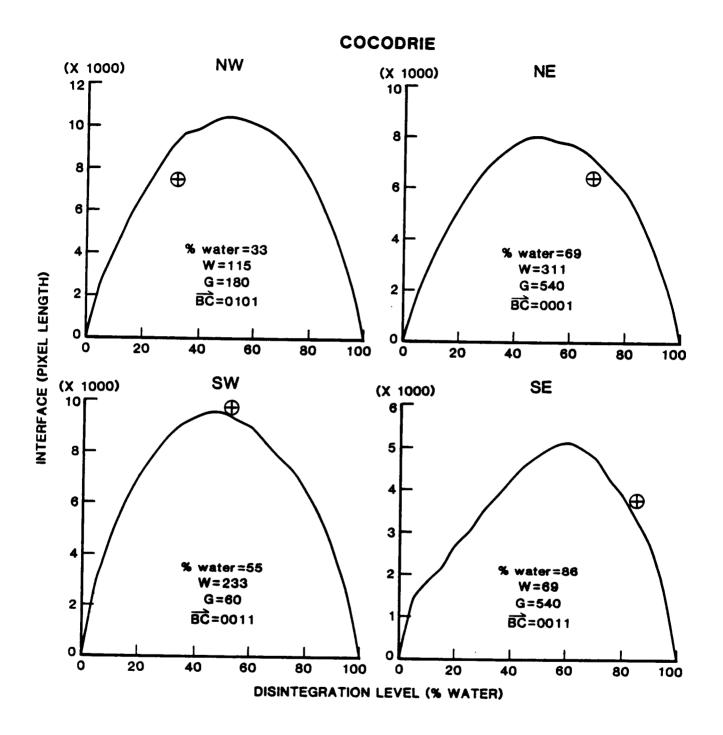


Figure D1-9. Simulated interface versus disintegration level, with TM scene coordinates, for the Cocodrie quadrangle.

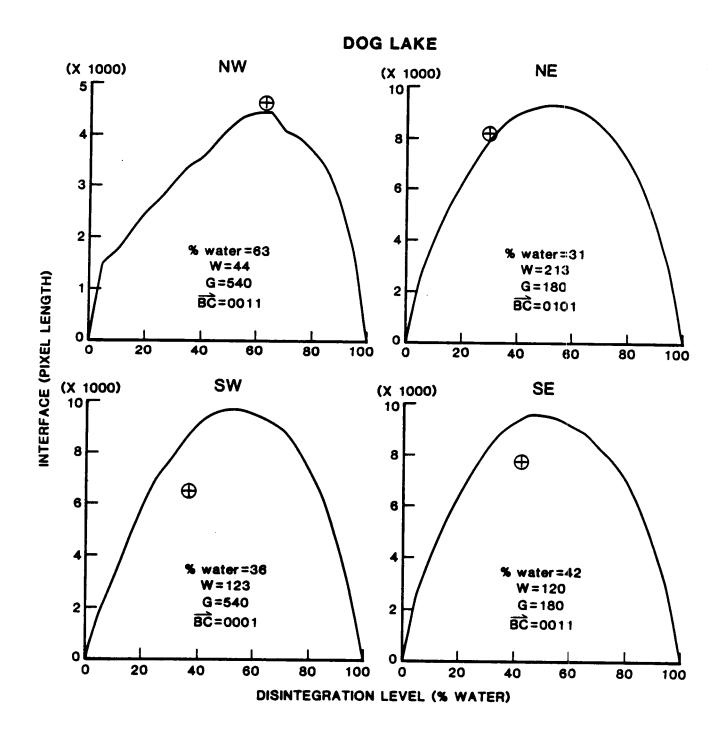


Figure D1-10. Simulated interface versus disintegration level, with TM scene coordinates, for the Dog Lake quadrangle.

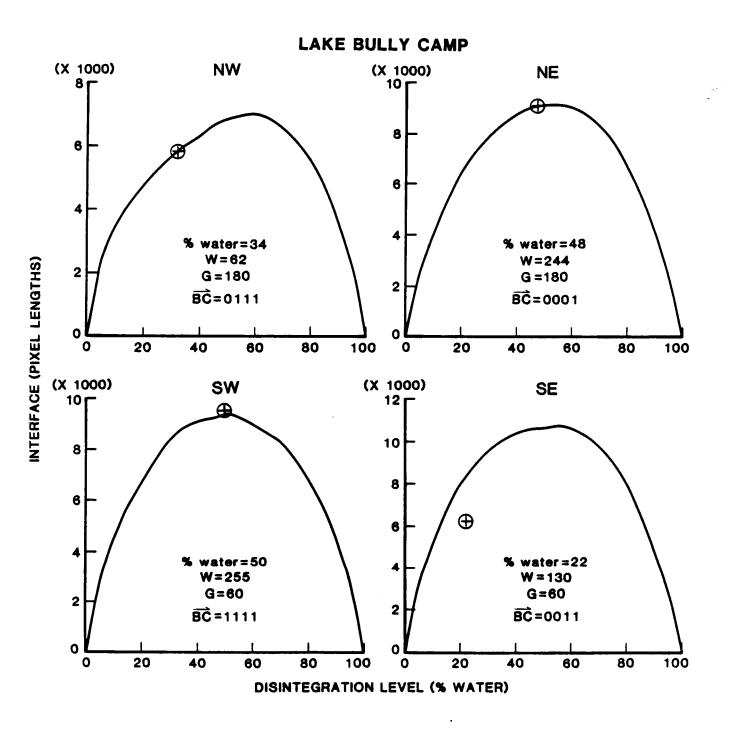


Figure D1-11. Simulated interface versus disintegration level, with TM scene coordinates, for the Lake Bully Camp quadrangle.

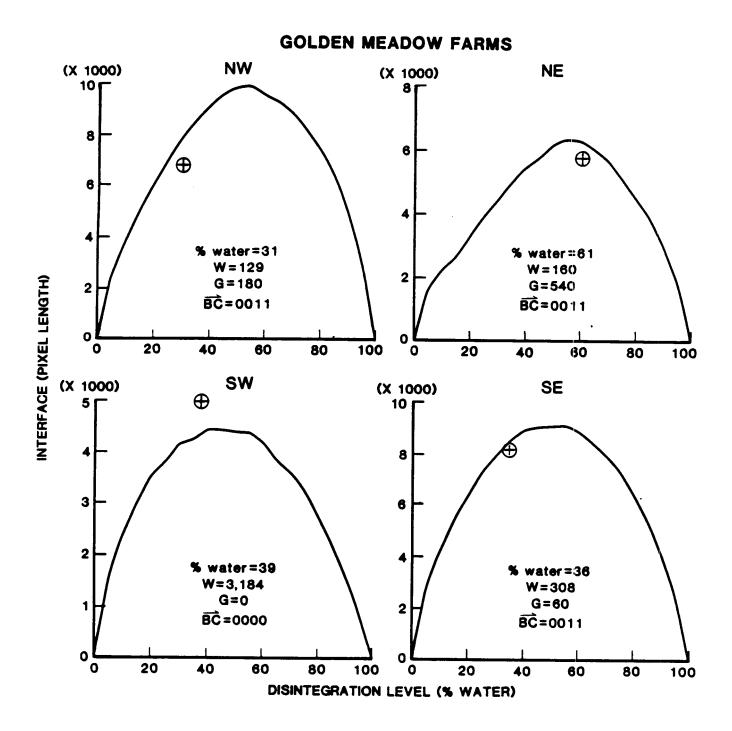


Figure D1-12. Simulated interface versus disintegration level, with TM scene coordinates, for the Golden Meadow Farms quadrangle.

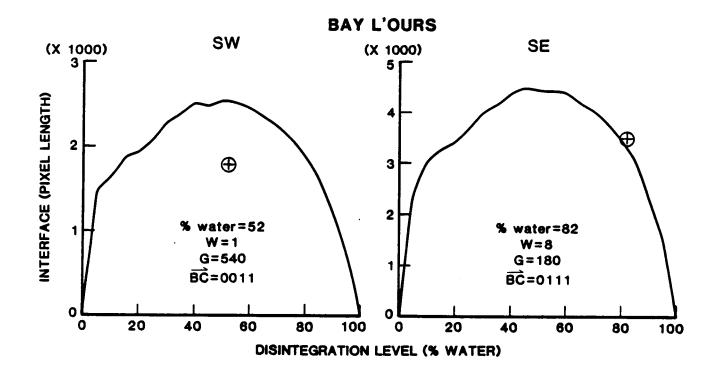


Figure D1-13. Simulated interface versus disintegration level, with TM scene coordinates, for the Bay L'Ours quadrangle.

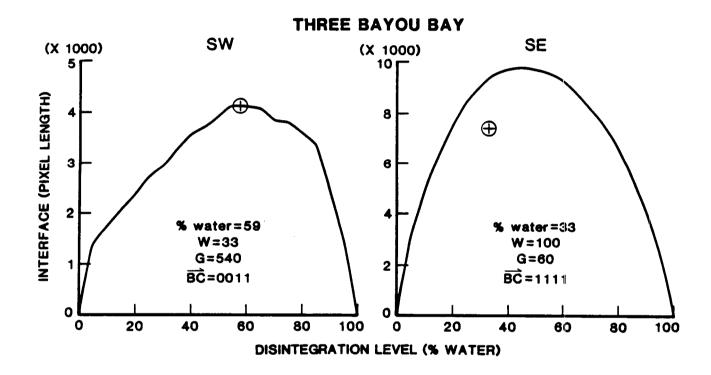


Figure D1-14. Simulated interface versus disintegration level, with TM scene coordinates, for the Three Bayou Bay quadrangle.

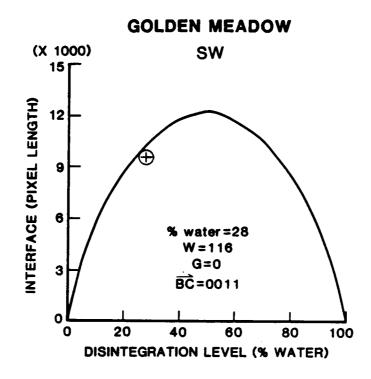


Figure D1-15. Simulated interface versus disintegration level, with TM scene coordinates, for the Golden Meadow quadrangle.

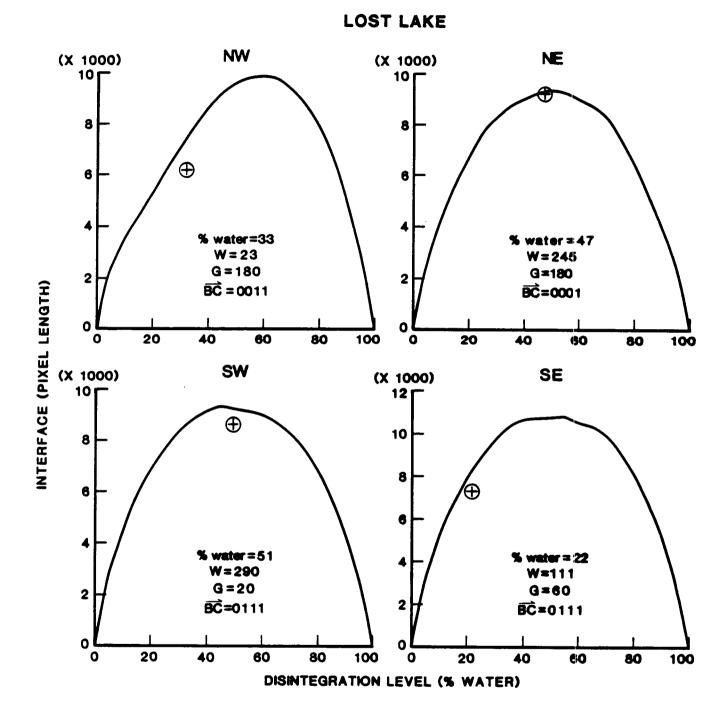


Figure D1-16. Simulated interface versus disintegration level, with TM scene coordinates, for the Lost Lake quadrangle.

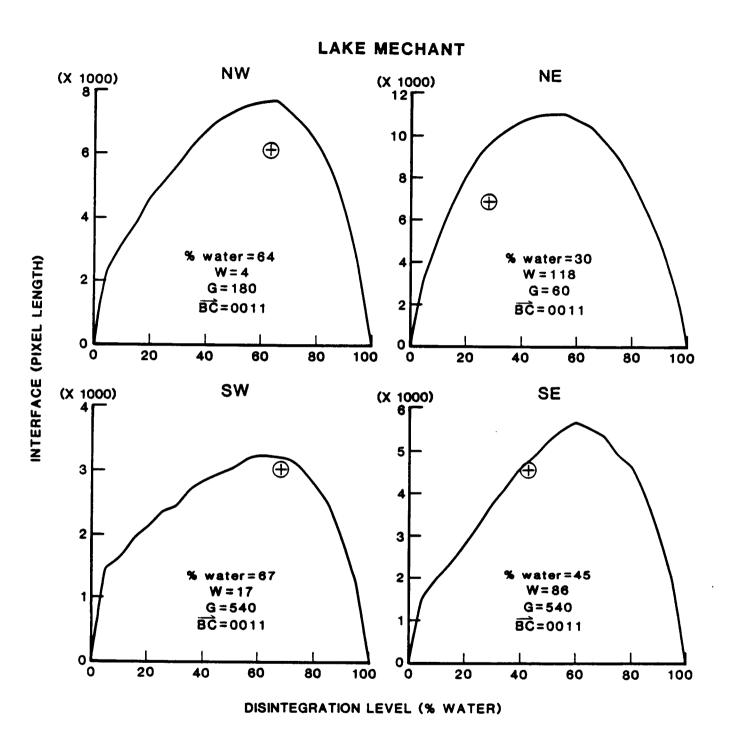


Figure D1-17. Simulated interface versus disintegration level, with TM scene coordinates, for the Lake Mechant quadrangle.

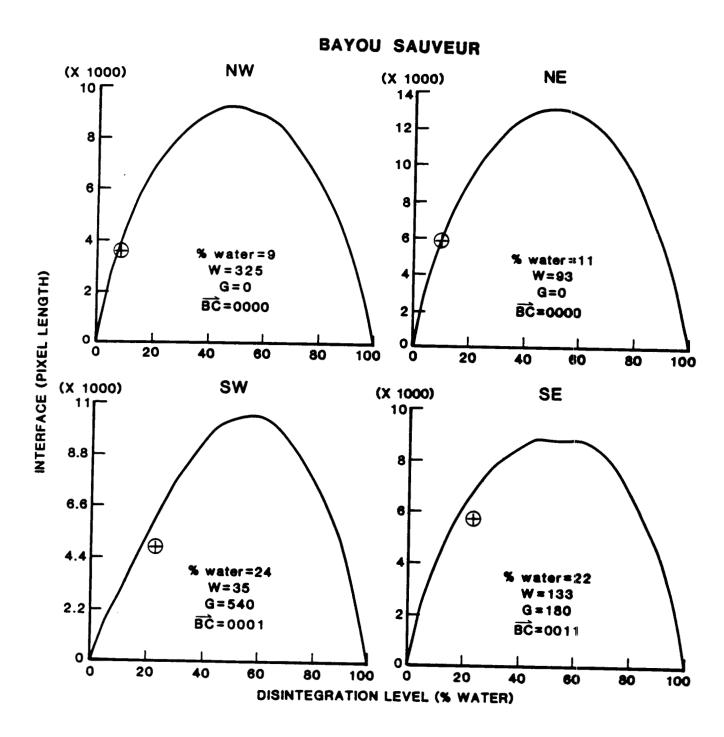


Figure D1-18. Simulated interface versus disintegration level, with TM scene coordinates, for the Bayou Sauveur quadrangle.

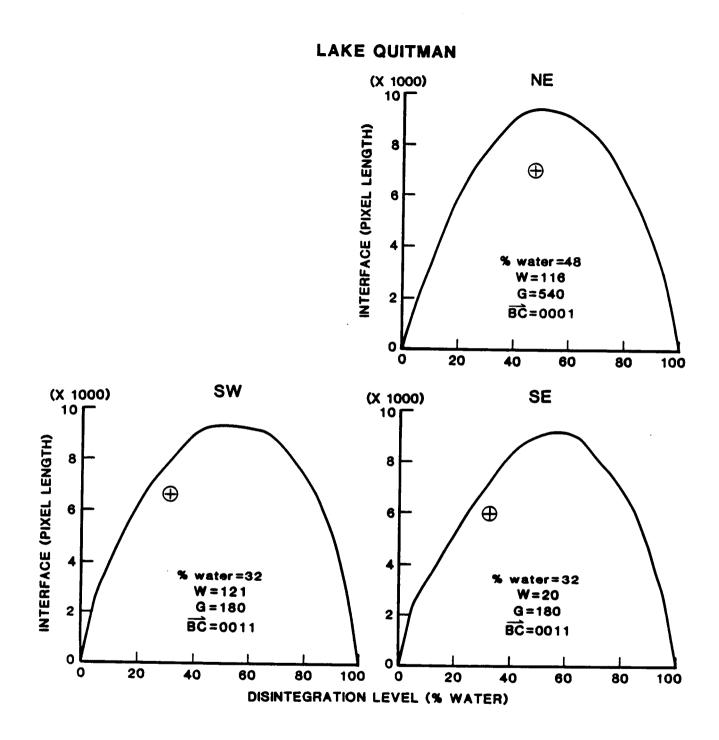


Figure D1-19. Simulated interface versus disintegration level, with TM scene coordinates, for the Lake Quitman quadrangle.

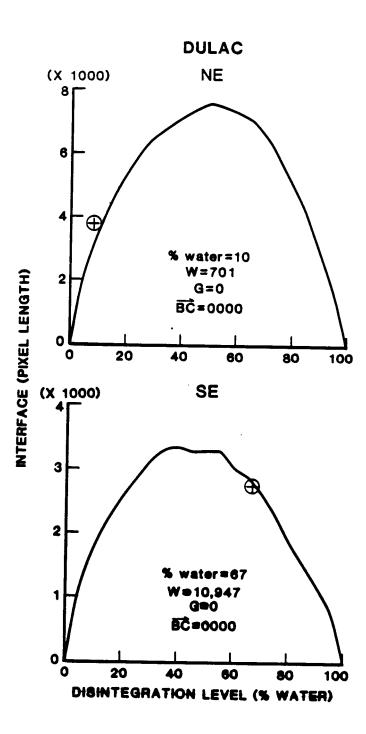


Figure D1-20. Simulated interface versus disintegration level, with TM scene coordinates, for the Dulac quadrangle.

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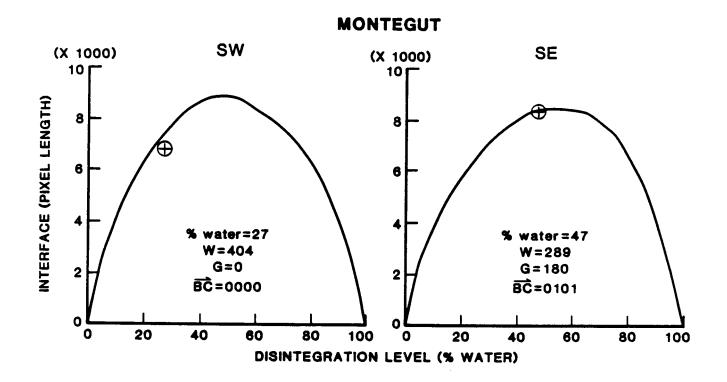


Figure D1-21. Simulated interface versus disintegration level, with TM scene coordinates, for the Montegut quadrangle.

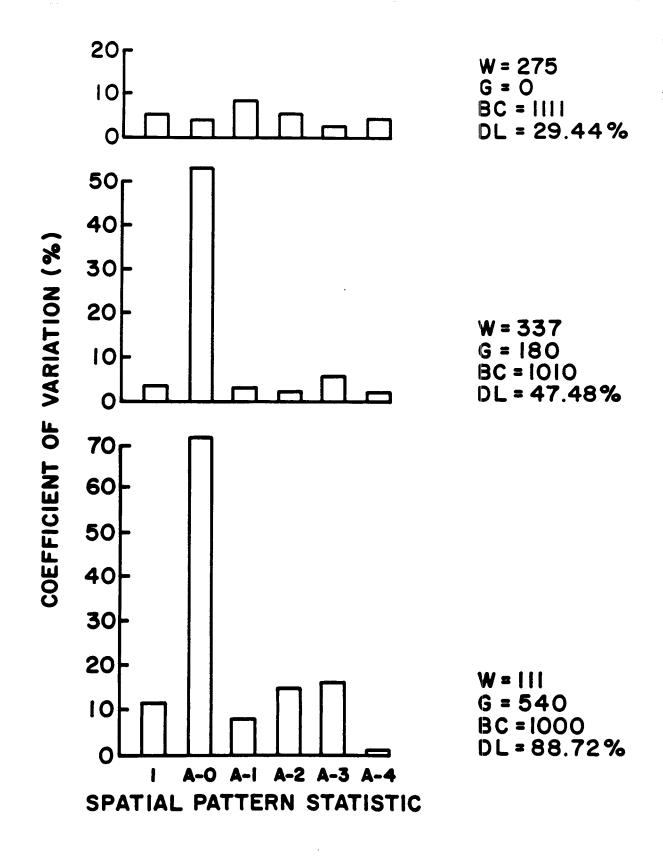


Figure D2. Coefficients of variation of spatial-pattern indices from 3 sets of 3 repeated simulations. The 3 repetitions of each set had the same W, G, BC and DL values. I = total interface; A-0 to A-4 refer to the number of water pixels in each 192 x 192 scene that had 0 to 4 sides adjacent to other water pixels

Appendix E

Comparison of Simulations with TM Scenes

- Table E1. TM image and simulation values for interface length, with deviation and percentage deviation.
- Table E2. TM image and simulation values for side-adjacency 0, with deviation and percentage deviation.
- Table E3. TM image and simulation values for side-adjacency 1, with deviation and percentage deviation.
- Table E4. TM image and simulation values for side-adjacency 2, with deviation and percentage deviation.
- Table E5. TM image and simulation values for side-adjacency 3, with deviation and percentage deviation.
- Table E6. TM image and simulation values for side-adjacency 4, with deviation and percentage deviation.
- Table E7. TM image and simulation values for border water pixels, with deviation and percentage deviation.
- Table E8. Results of simple linear regression of simulation spatial-pattern indices on corresponding TM-image spatial-pattern statistics.
- Figure E1. Comparisons of spatial patterns of marsh disintegration in TM scenes with "best-fit" simulations at the same disintegration level.
- Figure E2. Comparisons of pond size distributions of TM scenes with the "best-fit" simulations at the same disintegration level.
- Figure E3. Comparison of interface as a function of disintegration level for the 70 TM scenes and for simulations at the same disintegration level.
- Figure E4. TM imagery interface and side-adjacency statistics (number of pixels) compared with simulations at the same disintegration level.

			Interface	e Length	
Quad Qu	arter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafourc	he, sa	<u>lt</u>			
Leeville	NW	9,093	9,133	40	0.44
	NE	10,025	9,935	-90	-0.90
	SE	10,118	9,811	-307	-3.03
	SW	6,115	7,780	1,665	27.23
Mink Bayou	NW	7,166	7,241	75	1.05
	NE	5,519	6,359	840	15.22
	SE	8,420	8,681	261	3.10
	SW	7,937	8,163	226	2.85
Caminada Pas		6,778	6,989	211	3.11
	NE	1,341	1,323	-18	-1.34
	SE	101	190	89	88.12
	SW	3,884	3,946	62	1.60
Bay Tambour	NW	5,818	5,884	66	1.13
	NE	3,047	3,057	10	0.33
	SE	2,245	2,255	10	0.45
D-14 D	SW	3,515	3,774	259	7.37
Pelican Pass	NW	341	534	193	56.60
	NE	2,433	2,395	-38	-1.56
	SE Sw	2,100 385	2,357 442	257 57	12.24 14.81
Early Lafour	che, sa	alt			
Grand Bayou	NW	3,431	3,797	366	10.67
du Large	NE	3,621	4,048	427	11.79
-	SE	7,174	7,978	804	11.21
	SW	1,111	1,678	567	51.04
Lake La	NW	2,079	2,039	-40	-1.92
Graisse	NE	292	492	200	68.49
Central Isles	5 NW	3,439	3,548	109	3.17
Dernieres	NE	1,605	1,315	-290	-18.07
	SE	748	719	-29	-3.88
	SW	745	842	97	13.02
Cocodrie	NW	7,430	9,375	1,945	26.18
	NE	6,457	7,116	659	10.21
	SE	3,681	3,240	-441	-11.98
	SW	9,692	9,279	-413	-4.26
Dog Lake	NW	4,612	4,397	-215	-4.66
	NE	8,114	8,129	15	0.19
	SE	7,774	9,434	1,660	21.35
	SW	6,482	8,615	2,133	32.91

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Table E1. TM image and simulation values for interface length, with deviation and percentage deviation.

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			Interface	Length	
Quad Quar	ter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafourche	<u>, bra</u>	ckish			
Lake Bully	NW	5,689	5,831	142	2.50
Camp	NE	9,158	9,099	-59	-0.64
	SE	6,157	8,350	2,193	35.62
	SW	9,502	9,398	-104	-1.10
Golden Meadow	NW	6,870	8,050	1,180	17.18
Farms	NE	5,700	6,239	539	9.46
	SE	8,163	8,567	404	4.95
	SW	4,954	4,430	524	-10.58
Bay L'Ours	SE	3,572	3,308	-264	-7.39
	SW	1,755	2,484	729	41.54
Three Bayou	SE	7,393	9,304	1,911	25.85
Bay	SW	4,136	4,108	-28	-0.68
Golden Meadow	SW	9,367	10,187	820	8.75
Early Lafourch	e, br	ackish			
Lost Lake	NW	6,093	7,570	1,477	24.24
	NE	9,209	9, 387	178	1.93
	SE	7,227	8,247	1,020	14.11
	SW	8,645	9,253	608	7.03
Lake Mechant	NW	6,070	7,697	1,627	26.80
	NE	6,782	9,797	3,015	44.46
	SE	4,570	4,844	274	6.00
	SW	3,050	3,223	173	5.67
Bayou Sauveur	NW	3,578	3,997	419	11.71
-	NE	5,847	6,316	469	8.02
	SE	5,580	6,737	1,157	20.74
	SW	4,892	6,319	1,427	29.17
Lake Quitman	NE	6,983	9,468	2,485	35.59
	SE	5,981	7,261	1,280	21.40
	SW	6,545	8,002	1,458	22.26
Dulac	NE	3,708	3,445	-263	-7.09
	SE	2,710	2,853	143	5.28
Montegut	SE	8,315	8,551	236	2.84
inite gat	SW	6,787	7,439	652	9.61

Table E1. (cont.)

<u> </u>			Side-Adja	icency 0	
Quad Qu	arter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafourc	he, sal	<u>.t</u>			
Leeville	NW	114	21	-93	-81.58
	NE	159	17	-142	-89.31
	SE	146	16	-130	-89.04
	SW	54	353	299	553.70
Mink Bayou	NW	101	658	-74	-73.27
	NE	117	237	120	102.56
	SE	178	18	-160	-89.89
	SW	103	18	-85	-82.52
Caminada Pas		80	33	-47	-58.75
	NE	3	76	73	24.33
	SE	0	12	12	0.00
	SW	114	106	-8	-7.02
Bay Tambour	NW	61	111	50	81.97
	ŇE	18	74	56	311.11
	SE	24	7	-17	-70.83
n 1/ n	SW	28	70	42	150.00
Pelican Pass		1	1	0	0.00
	NE	7	11	4	57.14
	SE Sw	13 0	45 7	32 7	246.15 0.00
Early Lafour	<u>che, sa</u>	<u>1t</u>			
Grand Bayou	NW	48	114	66	137.50
du Large	NE	37	73	36	97.30
	SE	109	298	189	173.39
_	SW	17	266	249	14.65
Lake La	NW	17	2	-15	-88.24
Graisse	NE	0	1	1	0.00
Central Isle		31	58	27	87.10
Dernieres	NE	6	19	13	216.67
	SE	10	11	1	10.00
	SW	3	17	14	466.67
Cocodrie	NW	106	40	-66	-62.26
	NE	40	4	-36	-90.00
	SE	32	10	-22	-68.75
	SW	69	7	-62	-89.86
Dog Lake	NW	29	57	28	96.55
	NE	113	25	-88	-77.88
	SE	95	26	-69	-72.63
	SW	67	45	-22	-32.84

Table E2. TM image and simulation values for side-adjacency 0, with deviation and percentage deviation.

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			Side-Adja	cency 0	
Quad Quar	ter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafourche	<u>, bra</u>	<u>ckish</u>			
Lake Bully	NW	46	71	25	54.35
Camp	NE	101	9	-92	-91.09
	SE	122	44	-78	-63.93
	SW	78	13	-65	-83.33
Golden Meadow	NW	81	41	-40	-49.38
Farms	NE	80	20	-60	-75.00
	SE	86	19	-67	-77.91
	SW	47	3	-44	-93.62
Bay L'Ours	SE	42	107	65	154.76
	SW	34	386	352	1,035.29
Three Bayou	SE	108	34	-74	-68.52
Bay	SW	60	85	25	41.67
Golden Meadow	SW	182	45	-137	-75.27
Early Lafourch	e, br	<u>ackish</u>			
Lost Lake	NW	62	199	137	220.97
	NE	107	16	-91	-85.05
	SE	126	54	-72	-57.14
	SW	75	10	-65	-86.67
Lake Mechant	NW	69	412	343	497.10
	NE	90	39	-51	-56.67
	SE	37	67	30	81.08
	SW	23	134	111	482.61
Bayou Sauveur	NW	82	24	-58	-70.73
-	NE	160	69	-91	-56.88
	SE	79	45	-34	-43.04
	SW	36	148	112	311.11
Lake Quitman	NE	69	24	-45	-65.22
	SE	64	252	188	293.75
	SW	63	36	-27	-42.86
Dulac	NE	82	10	-72	-87.81
	SE	30	1	-29	-96.67
Montegut	SE	112	19	-93	-83.04
	SW	67	14	-53	-79.10

Table E2. (cont.)

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		· · · · · · · · · · · · · · · · · · ·	Side-adja	cency 1	
Quad Qu	arter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafouro	che, sal	. <u>t</u>			
Leeville	NW	439	665	226	51.48
	NE	713	802	89	12.48
	SE	519	760	241	46.44
	SW	253	1,174	921	364.03
Mink Bayou	NW	403	658	255	63.28
	NE	343	904	561	163.56
	SE	611	737	126	20.62
	SW	489	698	209	42.74
Caminada Pas	s NW	344	667	323	93.90
	NE	20	131	111	555.00
	SE	1	11	10	1,000.00
	SW	254	538	284	111.81
Bay Tambour	NW	233	747	514	220.60
-	NE	92	433	341	370.65
	SE	73	164	91	124.66
	SW	106	448	342	322.64
Pelican Pass		11	34	23	209.09
	NE	93	206	113	121.51
	SE	64	302	238	371.87
	SW	4	40	36	900.00
Early Lafour	<u>che, sa</u>	lt			
Grand Bayou	NW	114	510	396	347.37
du Large	NE	159	534	376	237.97
	SE	477	1,077	600	125.79
	SW	50	83	33	66.00
Lake La	NW	54	148	94	174.07
Graisse	NE	3	22	19	633.33
Central Isle	es NW	131	382	251	191.60
Dernieres	NE	33	143	110	333.33
	SE	18	73	55	305.56
	SW	12	74	62	516.67
Cocodrie	NW	493	846	353	71.60
	NE	234	435	201	85.90
	SE	153	229	76	49.67
	SW	410	599	189	46.10
	NW	159	480	321	201.89
Dog Lake					
род цаке	NE	449	710	261	58.13
Dog Lake	NE SE	449 403	710 798	261 395	58.13 98.01

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Table E3. TM image and simulation values for side adjacency 1, with deviation and percentage deviation.

				Side-adja	cency 1	
Quad	Quar	ter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Laf	fourche	e, bra	nckish			
Lake Bul	lly	NW	276	678	402	145.65
Camp		NE	562	662	100	17.79
		SE	413	805	392	94.92
		SW	469	676	207	44.14
Golden M	leadow	NW	414	761	347	83.82
Farms		NE	367	512	145	39.51
		SE	515	683	168	32.62
		SW	243	319	76	31.28
Bay L'Ou	ITS	SE	254	443	189	74.41
		SW	106	138	32	30.19
Three Ba	ayou	SE	469	831	362	77.19
Bay		SW	218	533	316	145.62
Golden M	leadow	SW	709	909	200	28.21
Early La	afourch	ie, br	ackish			
Lost Lak	ĸe	NW	329	1,088	759	230.70
		NE	550	672	122	22.18
		SE	496	821	325	65.52
		SW	442	620	178	40.27
Lake Med	chant	NW	325	1,137	812	249.85
		NE	379	857	478	126.12
		SE	168	526	358	213.10
		SW	122	440	318	260.66
Bayou Sa	auveur	NW	291	430	139	47.77
-		NE	556	776	220	39.57
		SE	323	678	355	109.91
		SW	199	903	704	353.77
Lake Qui	itman	NE	348	773	425	122.13
		SE	302	989	687	227.48
		SW	365	797	432	118.36
Dulac		NE	298	311	13	4.36
		SE	115	182	67	58.26
Montegut	2	SE	479	670	191	39.88
		SW	423	578	155	36.64
Mean			290	561	271	172.33
Mean			290	561	271	

Table E3. (cont.)

		Side-adjacency 2				
Quad Qu	arter	Image	Sim.	Deviation	<pre>% Deviation</pre>	
Late Lafourc	he, sal	<u>.t</u>				
Leeville	NW	2,124	1,853	-271	-12.75	
	NE	2,315	2,048	-267	-11.53	
	SE	2,470	1,947	-523	-21.17	
	SW	1,495	1,026	-469	-31.17	
Mink Bayou	NW	1,686	1,455	-231	-13.70	
	NE	1,316	969	-347	-26.37	
	SE	1,873	1,743	-130	-6.94	
	SW	1,867	1,635	-232	-12.43	
Caminada Pass		1,669	1,396	-273	-16.36	
	NE	418	169	-249	-59.57	
	SE	34	21	-13	-38.24	
	SW	952	653	-299	-31.41	
Bay Tambour	NW	1,506	1,013	-493	-32.74	
	NE	773	478	-295	-38.16	
	SE	539	451	-88	-16.33	
	SW	833	669	-164	-19.69	
Pelican Pass		67	82	15	22.39	
	NE	579	482	-97	-16.75	
	SE	500	418	-82	-16.40	
	SW	100	80	-20	-20.00	
Early Lafour	che, sa	<u>lt</u>				
Grand Bayou	NW	845	618	-227	-26.86	
du Large	NE	886	701	-185	-20.88	
	SE	1,688	1,277	-411	-24.35	
	SW	254	. 89	-165	-64.96	
Lake La	NW	475	413	-62	-13.05	
Graisse	NE	67	57	-10	-14.93	
					17 31	
Central Isle		809	669	-140	-17.31	
	s NW	809 432	669 224	-140 -208		
Central Isle	s NW NE	432	224	-208	-48.15	
Central Isle	S NW NE SE	432 186	22 4 120	-208 -66	-48.15 -35.48	
Central Isle Dernieres	S NW NE SE SW	432 186 173	224 120 163	-208 -66 -10	-48.15 -35.48 5.78	
Central Isle	s NW NE SE SW NW	432 186 173 1,692	224 120 163 1,949	-208 -66 -10 257	-48.15 -35.48 5.78 15.19	
Central Isle Dernieres	s NW NE SE SW NW NE	432 186 173 1,692 1,454	224 120 163 1,949 1,397	-208 -66 -10 257 -57	-48.15 -35.48 5.78 15.19 -3.92	
Central Isle Dernieres	S NW NE SE SW NW NE SE	432 186 173 1,692 1,454 813	224 120 163 1,949 1,397 659	-208 -66 -10 257 -57 -154	-48.15 -35.48 5.78 15.19 -3.92 -18.94	
Central Isle Dernieres Cocodrie	S NW NE SE SW NW NE SE SW	432 186 173 1,692 1,454 813 2,256	224 120 163 1,949 1,397 659 1,822	-208 -66 -10 257 -57 -154 -434	-48.15 -35.48 5.78 15.19 -3.92 -18.94 -19.24	
Central Isle Dernieres	S NW NE SE SW NW NE SE SW NW	432 186 173 1,692 1,454 813 2,256 1,084	224 120 163 1,949 1,397 659 1,822 869	-208 -66 -10 257 -57 -154 -434 -215	-48.15 -35.48 5.78 15.19 -3.92 -18.94 -19.24 -19.83	
Central Isle Dernieres Cocodrie	S NW NE SE SW NW NE SE SW	432 186 173 1,692 1,454 813 2,256	224 120 163 1,949 1,397 659 1,822	-208 -66 -10 257 -57 -154 -434	-48.15 -35.48 5.78 15.19 -3.92 -18.94 -19.24	

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Table E4. TM image and simulation values for side-adjacency 2, with deviation and percentage deviation.

		Side-adjacency 2				
Quad Qua	rter	Image	Sim.	Deviation	<pre>% Deviation</pre>	
Late Lafourch	e, bra	ckish				
Lake Bully	NW	1,259	1,126	-133	-10.56	
Camp	NE	2,039	1,800	-239	-11.72	
	SE	1,363	1,732	369	27.07	
	SW	2,124	1,845	-279	-13.14	
Golden Meadow		1,530	1,691	161	10.52	
Farms	NE	1,181	1,258	77	6.52	
	SE	1,813	1,751	-62	-3.42	
	SW	1,085	853	-232	-21.38	
Bay L'Ours	SE	820	517	-303	-36.95	
	SW	387	123	-264	-68.22	
Three Bayou	SE	1,670	1,887	217	12.99	
Bay	SW	974	749	-225	-23.10	
Golden Meadow	SW	2,118	2,078	-40	-1.89	
Early Lafourc	he, br	ackish				
Lost Lake	NW	1,352	1,227	-125	-9.25	
	NE	2,069	1,863	-206	-9.96	
	SE	1,644	1,677	33	2.01	
	SW	1,912	1,853	-59	-3.09	
Lake Mechant	NW	1,449	976	-473	-32.64	
	NE	1,613	1,998	385	23.87	
	SE	1,197	946	-251	-20.97	
	SW	823	474	-349	-42.41	
Bayou Sauveur		771	820	49	6.36	
-	NE	1,201	1,244	43	3.58	
	SE	1,309	1,330	21	1.60	
	SW	1,245	1,067	-178	-14.30	
Lake Quitman	NE	1,608	1,880	272	16.92	
	SE	1,421	1,167	-254	-17.88	
	SW	1,574	1,577	3	0.19	
Dulac	NE	791	708	-83	-10.49	
	SE	666	502	-164	-24.63	
Montegut	SE	1,828	1,674	-154	-8.43	
multicegae	SW	1,521	1,577	56	3.68	
Mean		1,226	1,098	-127	-14.28	

Table E4. (cont.)

		Side-adjacency 3					
Quad Q	uarter	Image	Sim.	Deviation	<pre>% Deviation</pre>		
Late Lafour	che, sal	<u>t</u>	**************************************				
Leeville	NW	2,964	3,295	331	11.17		
	NE	2,473	3,279	806	32.59		
	SE	2,901	3,440	539	18.58		
	SW	2,061	721	-1,340	-65.02		
Mink Bayou	NW	2,067	2,218	151	7.31		
	NE	1,289	730	-559	-43.37		
	SE	1,977	2,845	868	43.91		
	SW	2,192	2,678	486	22.17		
Caminada Pa	ss NW	1,998	2,041	43	2.15		
	NE	411	280	-131	-31.87		
	SE	25	53	28	112.00		
	SW	705	582	-123	-17.45		
Bay Tambour	NW	1,788	1,153	-635	-35.52		
	NE	1,112	471	-641	-57.64		
	SE	826	831	5	0.61		
	SW	1,364	796	-568	-41.64		
Pelican Pass	s NW	161	261	100	62.11		
	NE	932	763	-169	-18.13		
	SE	839	411	-428	-51.01		
	SW	160	134	-26	-16.25		
Early Lafou	rche, sa	lt					
Grand Bayou	NW	1,151	542	-609	-52.91		
du Large	NE	1,145	726	-419	-36.59		
	SE	1,839	937	-902	-49.05		
	SW	367	158	-209	-56.95		
Lake La	NW	853	748	-105	-12.31		
Graisse	NE	143	302	159	111.19		
Central Isl	es NW	1,268	832	-436	-34.39		
Dernieres	NE	601	360	-241	-40.10		
	SE	258	210	-48	-18.61		
	SW	334	226	-108	-32.34		
Cocodrie	NW	2,005	2,741	736	36.17		
	NE	2,596	2,931	335	12.90		
	SE	1,387	1,165	-222	-16.01		
	SW	3,511	3,758	247	7.04		
		1,769	954	-815	-46.07		
Dog Lake	NW	1,703	221				
Dog Lake	NW NE	2,394	2,542	148	6.18		
Dog Lake							

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Table E5. TM image and simulation values for side-adjacency 3, with deviation and percentage deviation.

		Side-adjacency 3				
Quad Quar	ter	Image	Sim.	Deviation	<pre>% Deviation</pre>	
Late Lafourche	e, bra	ckish				
Lake Bully	NW	2,079	1,246	-833	-40.07	
Camp	NE	2,845	3,402	557	19.58	
	SE	1,578	2,224	646	40.94	
	SW	3,373	3,610	237	7.03	
Golden Meadow	NW	2,179	2,186	7	0.32	
Farms	NE	1,836	2,072	236	12.85	
	SE	2,538	2,872	334	13.16	
	SW	1,775	1,731	-44	-2.48	
Bay L'Ours	SE	963	503	-460	-47.77	
	SW	513	269	-244	-47.56	
Three Bayou	SE	2,108	2,887	779	36.95	
Bay	SW	1,218	664	-554	-45.49	
Golden Meadow	SW	2,184	3,029	845	38.69	
Early Lafourch	ne, br	ackish				
Lost Lake	NW	2,074	1,008	1,056	-51.40	
	NE	2,848	3,512	664	23.32	
	SE	1,820	2,152	332	18.24	
	SW	3,078	3,545	467	15.17	
Lake Mechant	NW	1,836	594	-1,242	-67.65	
20110 1100110110	NE	1,957	3,003	1,046	53.45	
	SE	1,471	1,089	-382	-25.97	
	SW	894	394	-550	-55.93	
Bayou Sauveur	NW	791	942	151	19.09	
	NE	1,030	1,139	109	10.58	
	SE	1,559	1,833	274	17.58	
	SW	1,592	829	-763	-47.93	
Lake Quitman	NE	2,326	3,208	882	37.92	
	SE	1,881	897	-984	-52.31	
	SW	1,948	2,283	335	17.20	
Dulac	NE	883	1,024	141	15.97	
JULUV	SE	854	1,265	411	48.13	
Montegut	SE	2,657	3,087	430	16.18	
noncegue	SW	2,103	2,417	314	14.93	
Mean		1,617	1,609	-8	-3.50	

Table E5. (cont.)

		Side-adjacency 4					
Quad Qu	arter	Image	Sim.	Deviation	<pre>% Deviation</pre>		
Late Lafourc	he, sa	<u>lt</u>					
Leeville	NW	9,856	9,551	-305	-3.10		
	NE	6,860	6,534	-326	-4.75		
	SE	8,369	8,299	-70	-0.84		
	SW	13,064	13,460	396	3.03		
Mink Bayou	NW	5,326	5,004	-322	-6.05		
	NE	5,744	5,778	34	0.59		
	SE	5,975	5,397	-578	-9.67		
	SW	6,026	5,584	-442	-7.34		
Caminada Pas	s NW	18,306	18,074	-232	-1.27		
	NE	29,050	29, 187	137	0.47		
	SE	35,792	35,763	-29	-0.08		
	SW	18,845	19,006	161	0.85		
Bay Tambour	NW	14,891	15,202	311	2.09		
•	NE	25,719	26,191	472	1.84		
	SE	30,541	30,542	1	0.00		
	SW	28,120	28,413	293	1.04		
Pelican Pass	NW	35,491	35,356	-135	-0.38		
	NE	28,221	28,283	62	0.22		
	SE	29,910	30,161	251	0.84		
	SW	34,833	34,822	-11	-0.03		
Early Lafour	che, s	alt					
Grand Bayou	NW	17,371	17,550	179	1.30		
du Large	NE	22,053	22,217	164	0.74		
	SE	10,349	10,845	496	4.79		
	SW	29,643	29,645	2	0.01		
Lake La	NW	31,517	31,565	48	0.15		
Graisse	NE	35,641	35,459	-182	-0.51		
Central Isle	s NW	25,018	25,227	209	0.84		
Dernieres	NE	31,419	31,723	304	0.97		
	SE	33,918	33,935	17	0.05		
	SW	33,245	33,236	-9	-0.03		
Cocodrie	NW	7,765	6,287	-1,478	-19.03		
	NE	20,598	20,194	-404	-1.96		
	SE	28,656	28,964	308	1.08		
		13,522	13,490	-32	-0.24		
	DW						
Dog Lake	sw Nw		20,321	533	2.69		
Dog Lake	NW	19,788	20,321 5,937	533 -281	2.69 -4.52		
Dog Lake			20,321 5,937 9,381	533 -281 -1,133	2.69 -4.52 -10.78		

Table E6. TM image and simulation values for side-adjacency 4, with deviation and percentage deviation.

		Side-adjacency 4					
Quad Qua	rter	Image	Sim.	Deviation	<pre>% Deviation</pre>		
Late Lafourch	e, bra	ackish					
Lake Bully	NW	8,482	8,680	198	2.33		
Camp	NE	11,744	11,405	-339	-2.89		
	SE	4,481	2,951	-1,530	-34.14		
	SW	11,989	11,537	-452	-3.77		
Golden Meadow	NW	6,931	6,281	-650	-9.38		
Farms	NE	18,357	17,993	-364	-1.98		
	SE	7,862	7,320	-542	-6.89		
	SW	11,049	11,360	311	2.82		
Bay L'Ours	SE	27,359	27,790	431	1.58		
	SW	17,837	17,852	15	0.08		
Three Bayou	SE	7,410	5,743	-1,667	-22.50		
Bay	SW	18,775	19,110	335	1.78		
Golden Meadow	SW	4,802	4,006	-796	-16.58		
Early Lafourc	he, b	rackish					
Lost Lake	NW	8,031	8,059	28	0.35		
	NE	11,576	11,034	-542	-4.68		
	SE	3,689	2,752	-937	-25.40		
	SW	12,874	12,138	-736	-5.72		
Lake Mechant	NW	19,400	19,940	540	2.78		
	NE	6,710	4,663	-2,047	-30.51		
	SE	13,303	13,462	159	1.20		
	SW	22,310	22,755	445	2.00		
Bayou Sauveur	NW	1,239	959	-280	-22.60		
	NE	966	714	-252	-26.09		
	SE	4,721	3,885	-836	-17.71		
	SW	5,748	5,730	-18	-0.31		
Lake Quitman	NE	13,105	11,475	-1,630	-12.44		
	SE	8,035	8,185	150	1.87		
	SW	7,490	6,598	-892	-11.91		
Dulac	NE	1,468	1,460	-8	-0.55		
	SE	22,739	22,312	-427	-1.88		
Montegut	SE	12,077	11,578	-499	-4.13		
-	SW	5,661	5,303	-358	-6.32		
Mean		35,792	35,763	-227	-4.44		

Table E6. (cont.)

				order water pi	XETZ
Quad Quar	ter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafourche	,_sal	t			
Leeville	NW	332	443	111	33.43
	NE	355	195	-160	-45.07
	SE	308	251	-57	-18.51
	SW	304	497	193	63.49
Mink Bayou	NW	219	439	220	100.46
	NE	236	426	190	80.51
	SE	237	110	-127	-53.59
	SW	225	289	64	28.44
Caminada Pass	NW	493	679	186	37.73
	NE	646	705	59	9.13
	SE	741	733	-8	-1.08
	SW	518	503	-15	-2.90
Bay Tambour	NW	388	640	252	64.95
	NE	504	570	66	13.10
	SE	704	711	7	0.99
	SW	654	708	54	8.26
Pelican Pass	NW	751	748	-3	-0.40
	NE	611	703	86	13.94
	SE	629	618	-11	-1.75
	SW	751	764	13	1.73
Early Lafourch	e, sal	lt			
Grand Bayou	NW	316	510	194	61.39
du Large	NE	514	541	27	5.25
-	SE	444	471	27	6.08
	SW	530	619	89	16.79
Lake La	NW	679	719	40	5.89
Graisse	NE	744	756	12	1.61
Central Isles	NW	675	764	89	13.19
Dernieres	NE	706	727	21	2.98
	SE	698	739	41	5.87
	SW	713	764	51	7.15
Cocodrie	NW	264	461	197	74.62
	NE	475	436	-39	-8.21
	SE	583	597	-39	2.40
	SW	402	493	91	22.64
		385	532	147	38.18
Dog Lake			JJ <u>4</u>	14/	JU.IO
Dog Lake	NW NF				
Dog Lake	NW NE SE	203 317	453 459	190 142	72.24 44.80

Table E7. TM image and simulation values for border water pixels, with deviation and percentage deviation.

		Num	ber of borde	er water pixe	ls
Quad Qua	arter	Image	Sim.	Deviation	<pre>% Deviation</pre>
Late Lafourc	he, bra	<u>ckish</u>			
Lake Bully	NW	276	617	341	123.55
Camp	NE	344	357	13	3.78
	SE	230	431	201	87.39
	SW	406	758	352	86.70
Golden Meado		261	435	174	66.67
Farms	NE	553	519	-34	-6.15
	SE	294	462	168	57.14
• -	SW	159	92	-67	-42.14
Bay L'Ours	SE	617	694	77	12.48
	SW	387	495	108	27.91
Three Bayou	SE	375	758	383	102.13
Bay	SW	501	501	0	0.00
Golden Meador	w SW	202	129	-73	-36.14
Early Lafour	che, br	ackish			
Lost Lake	NW	187	453	266	142.25
	NE	294	347	53	18.03
	SE	265	584	319	120.38
	SW	340	555	215	63.24
Lake Mechant	NW	534	554	20	3.75
	NE	267	464	197	73.78
	SE	380	465	85	22.37
	SW	550	524	-26	-4.73
Bayou Sauveu		42	41	-1	-2.38
	NE	106	76	-30	-28.30
	SE	203	422	219	107.88
	SW	140	283	143	102.14
Lake Quitman	NE	312	408	96	30.77
	SE	244	456	212	86.89
	SW	301	449	148	49.17
Dulac	NE	29	37	8	27.59
	SE	221	363	142	64.25
Montegut	SE	349	473	124	35.53
	SW	237	123	-114	-48.10
Mean		404	492	87	29.10

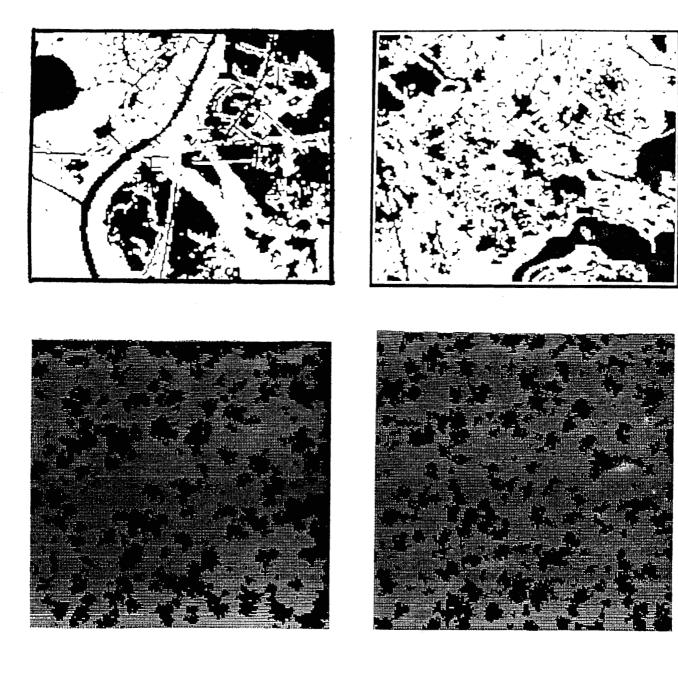
Table E7. (cont.)

Table E8. Results of simple linea pattern indices on con indices.	r regression of simulation spatial rresponding image spatial pattern
Interface_length	
Sim = 154.8 + 1.069 • Image (.4250) (.0001)	$R^2 = 94.12$ P of F-stat = .0001
<u>0-j54</u>	
Sim = 73.23 - 0.0739 • Image (.0005) (.7667)	R2 = .13 P of F-stat = .7667
<u>Adj-1</u>	
Sim = 212.4 + 1.204 • Image (.0001) (.0001)	$R^2 = 56.14$ P of F-stat = .0001
<u>Adj-2</u>	
Sim = -88.01 + 0.9680 • Image (.0929) (.0001)	R ² = 90.72 P of F-stat = .0001
<u>Adj-3</u>	
Sim = -325.7 + 1.213 • Image (.0113) (.0001)	$R^2 = 79.67$ P of F-stat = .0001
<u>Adj-4</u>	
$Sim = -640.3 + 1.026 \cdot Image$ (.0001) (.0001)	R ² = 99.77 P of F-stat = .0001

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Note: Values in parentheses are T-stat probability levels.

Figure E1. Comparisons of spatial patterns of marsh disintegration in TM scenes (upper images) with "best-fit" simulations at the same disintegration level (lower images). Quadrangles are presented in the same order as in Table D1.



W = 272 G = 60 DL = 42.94 BC = 0011



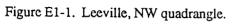


Figure E1-2. Lecville, NE quadrangle.

W = 188 G = 0 DL = 34.93 BC = 0111

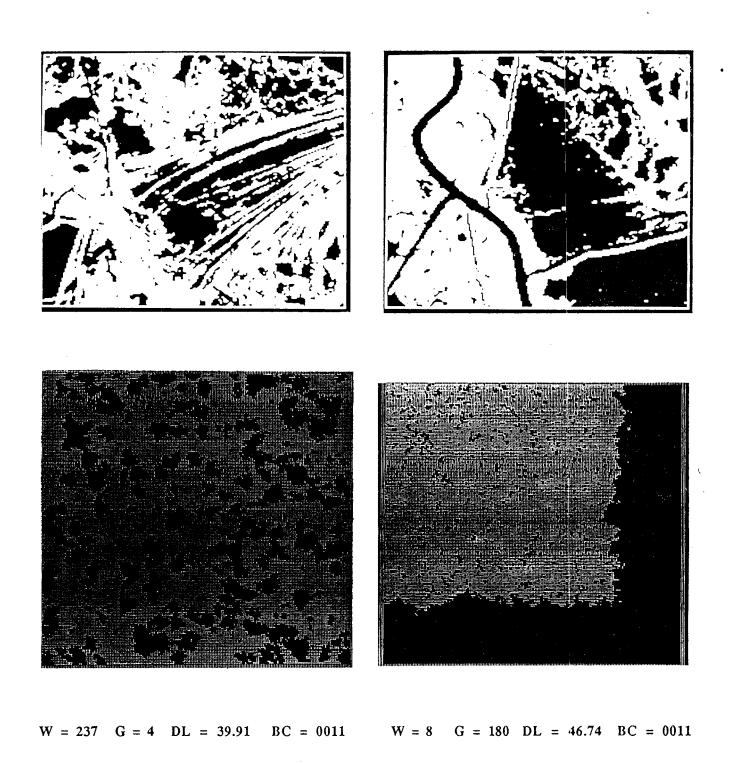
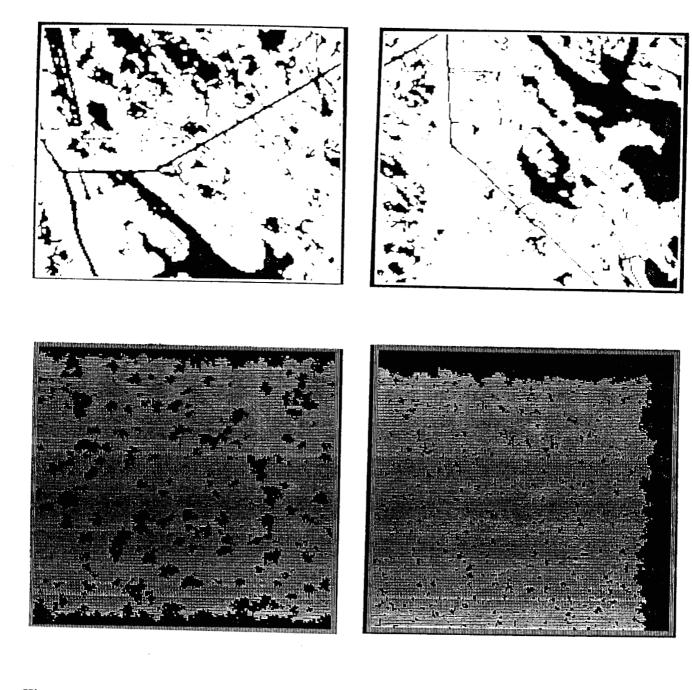
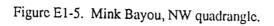


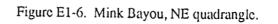
Figure E1-3. Leeville, SE quadrangle.

Figure E1-4. Leeville, SW quadrangle.

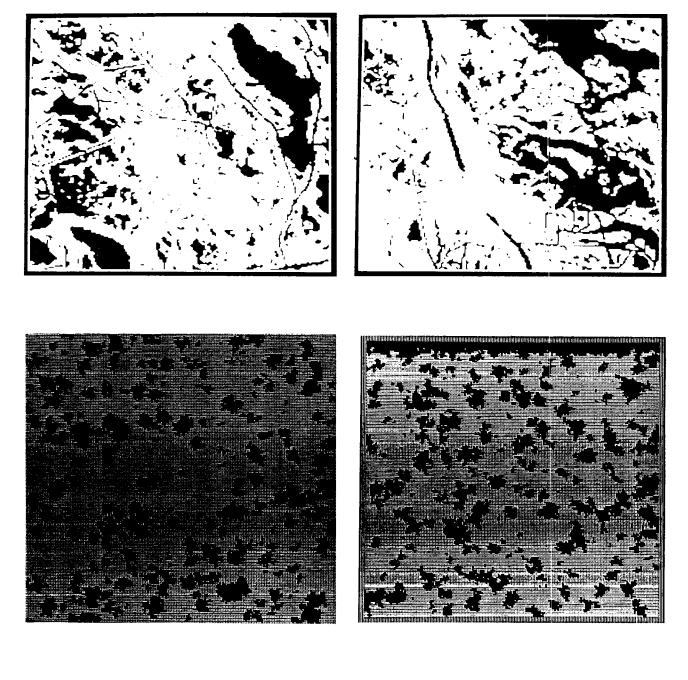


W = 244 G = 180 DL = 26.59 BC = 0101





W = 24 G = 180 DL = 24.54 BC = 0011



W = 248 G = 0 DL = 29.44 BC = 0111 W = 261 G = 180 DL = 29.57 BC = 0001

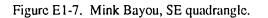
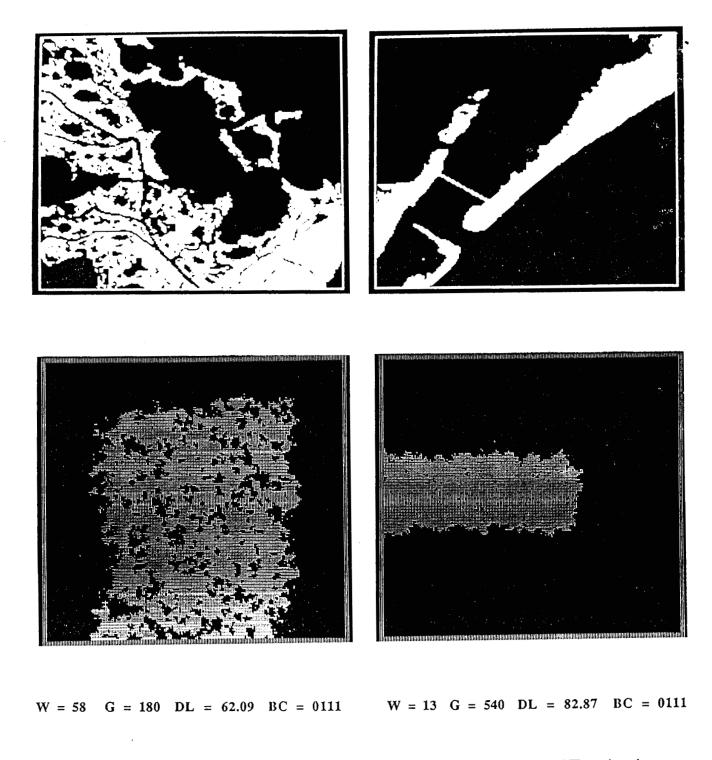
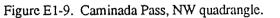
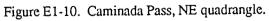


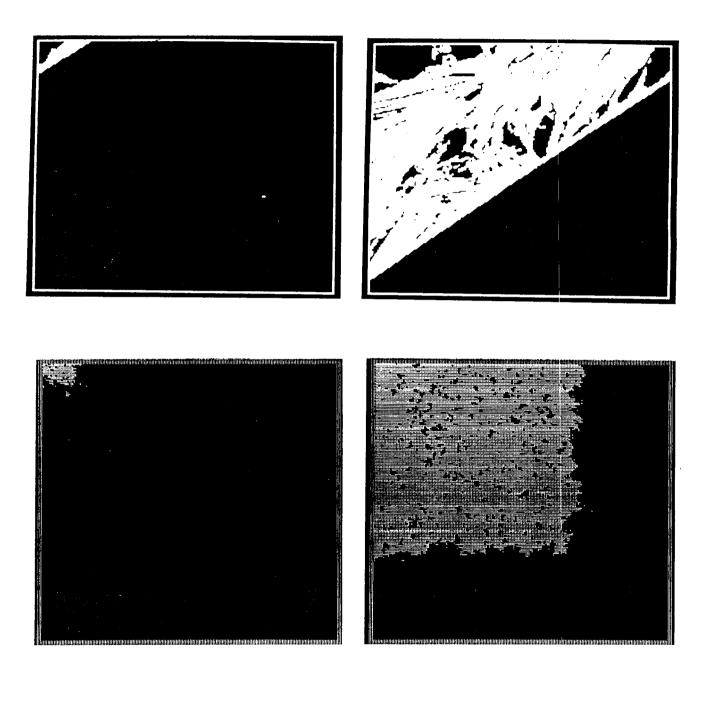
Figure E1-8. Mink Bayou, SW quadrangle.





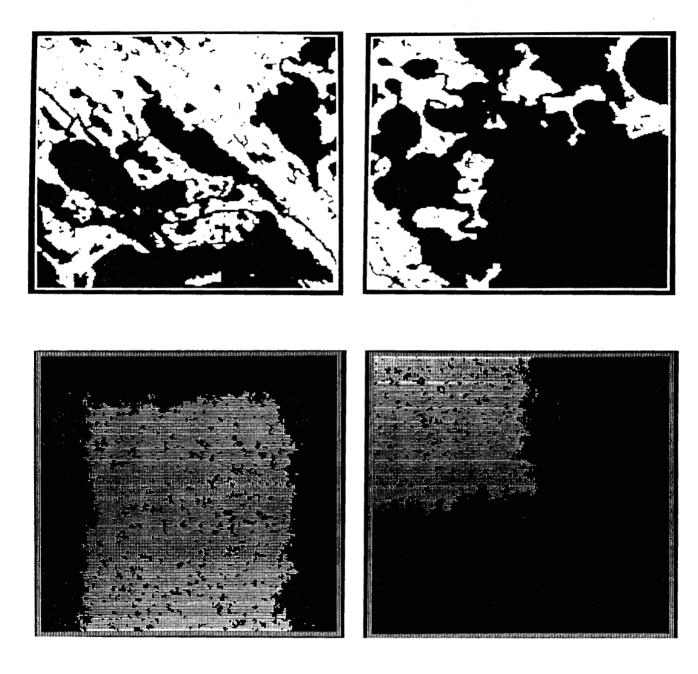
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W = 1DL = 99.27 BC = 0011 W = 28 G = 540 DL = 58.02 BC = 0011 $\mathbf{G} = \mathbf{540}$





W = 31 G = 180 DL = 51.18 BC = 0111 W = 17 G = 540 DL = 76.55 BC = 0011

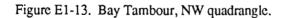


Figure E1-14. Bay Tambour, NE quadrangle.

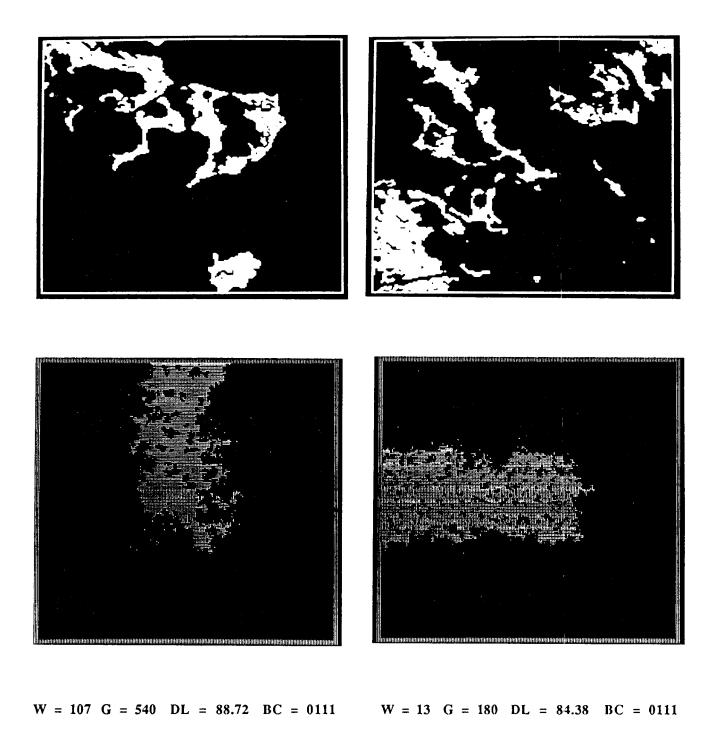
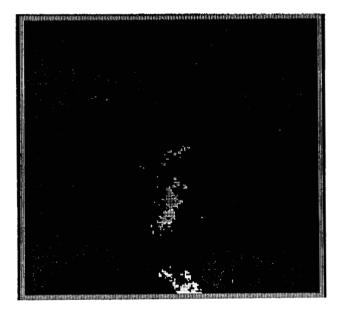


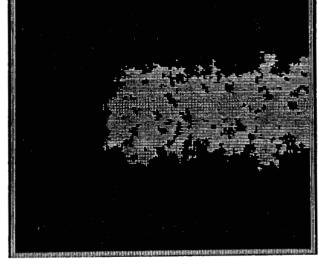
Figure E1-15. Bay Tambour, SE quadrangle. Figure E1-16. Bay Tambour, SW quadrangle.







W = 8 G = 180 DL = 98.96 BC = 0111





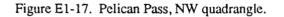
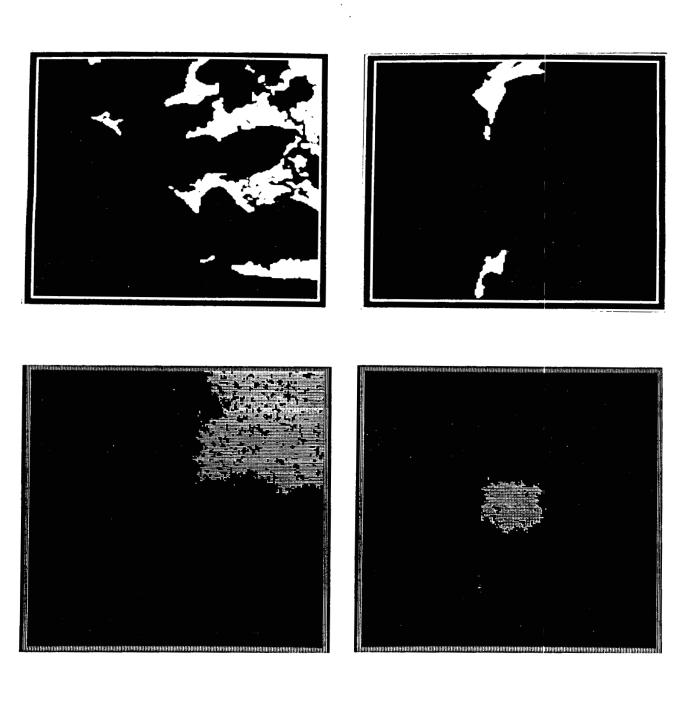


Figure E1-18. Pelican Pass, NE quadrangle.



W = 16 G = 540 DL = 86.68 BC = 0011 W = 38 G = 540 DL = 97.24 BC = 1111

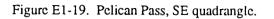
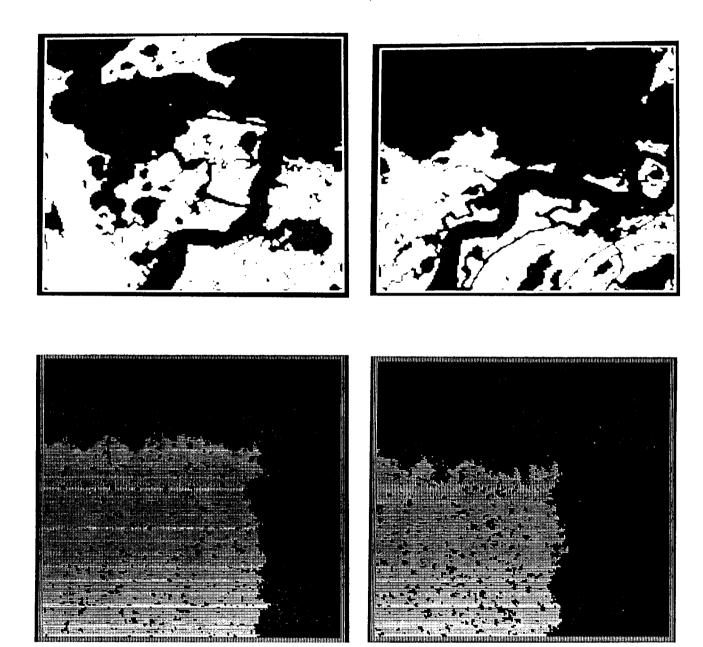
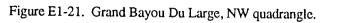


Figure E1-20. Pelican Pass, SW quadrangle.



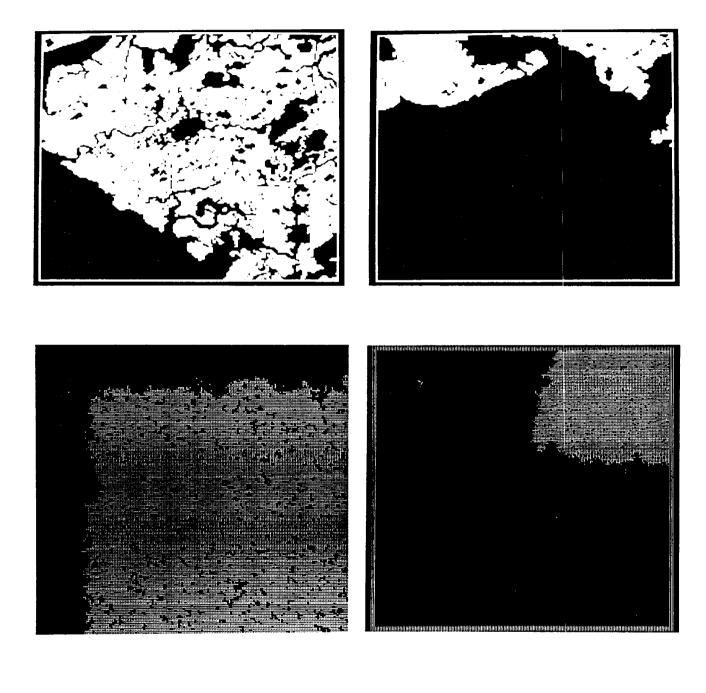
W = 30 G = 540 DL = 53.83 BC = 0011

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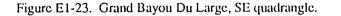


W = 29 G = 540 DL = 67.26 BC = 0011

Figure E1-22. Grand Bayou Du Large, NE quadrangle.

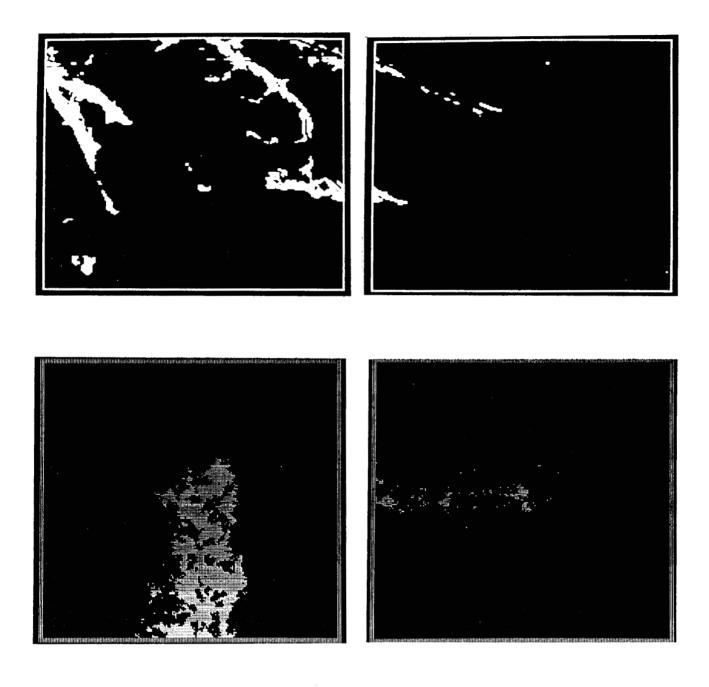


W = 14 G = 180 DL = 40.44 BC = 0011



W = 0 G = 540 DL = 83.72 BC = 0011

Figure E1-24. Grand Bayou Du Large, SW quadrangle.



W = 113 G = 540 DL = 91.13 BC = 0111



W = 10 G = 180 DL = 99.28 BC = 0111

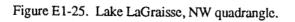


Figure E1-26. Lake LaGraisse, NE quadrangle.

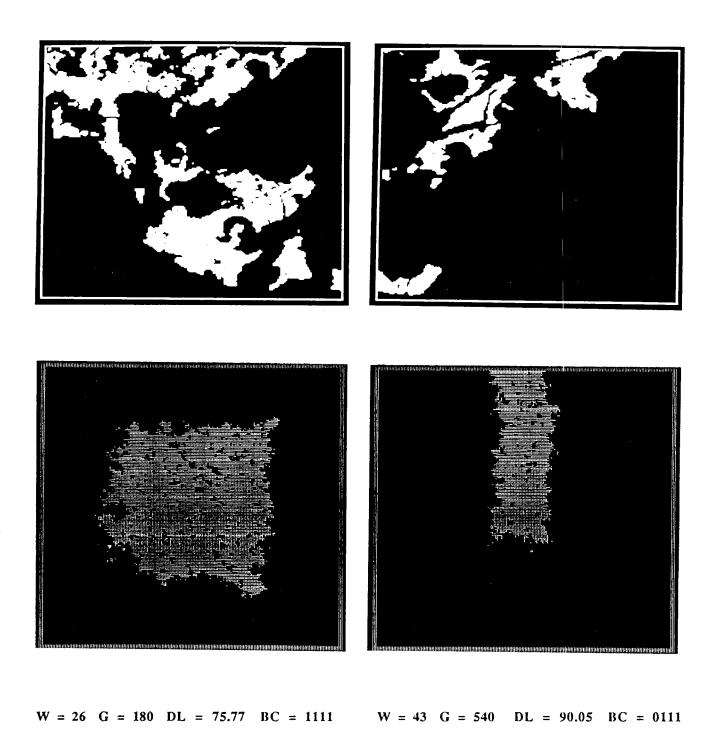
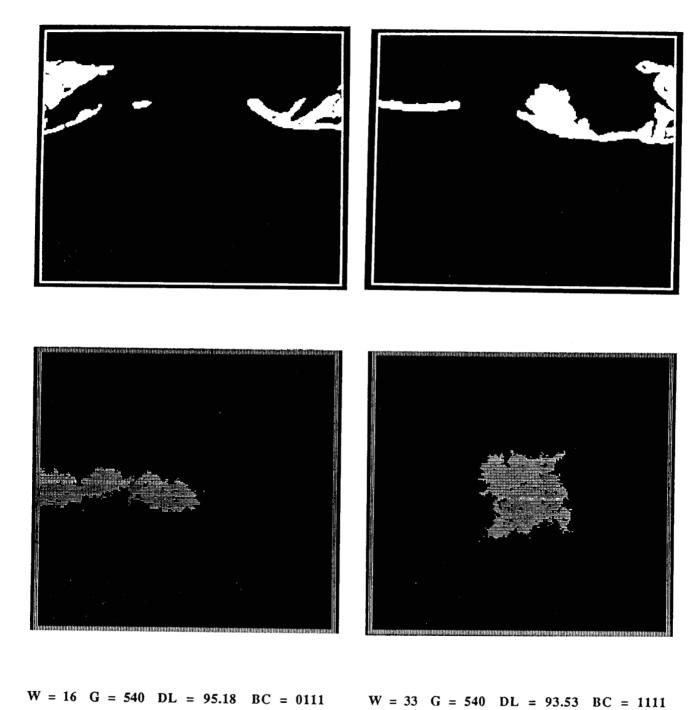


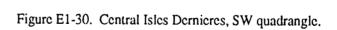
Figure E1-27. Central Isles Dernieres, NW quadrangle.

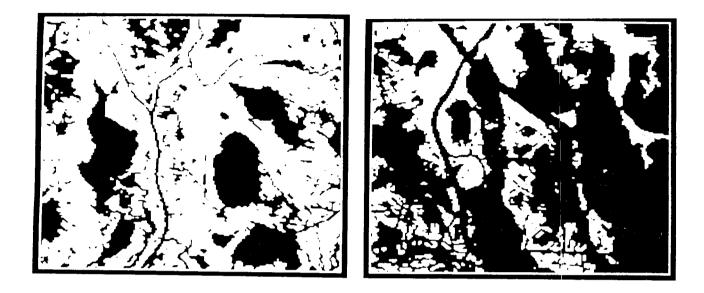
Figure E1-28. Central Isles Dernieres, NE quadrangle.



W = 16 G = 540 DL = 95.18 BC = 0111







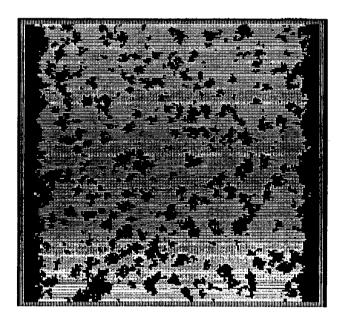
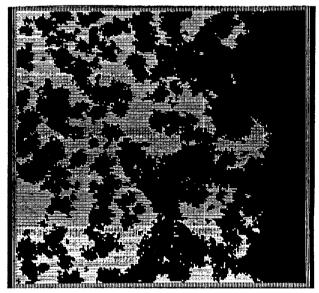




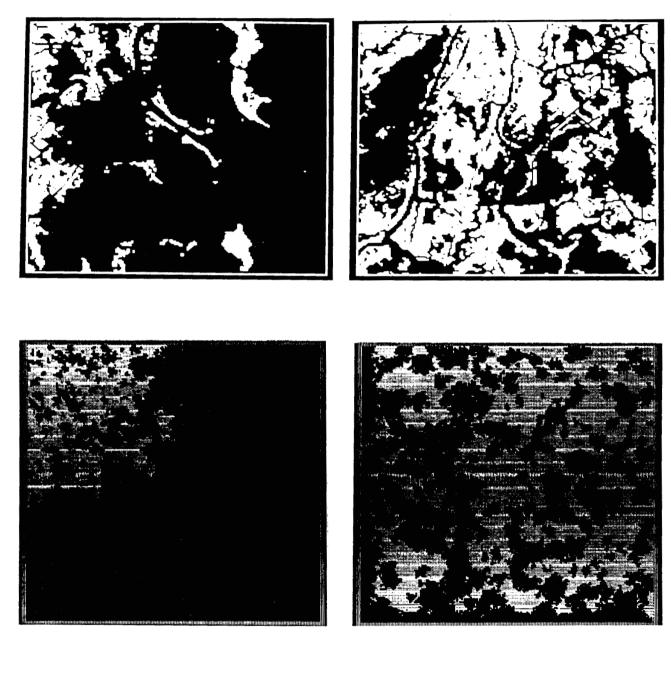
Figure E1-31. Cocodrie, NW quadrangle.



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Figure E1-32. Cocodric, NE quadrangle.





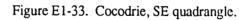
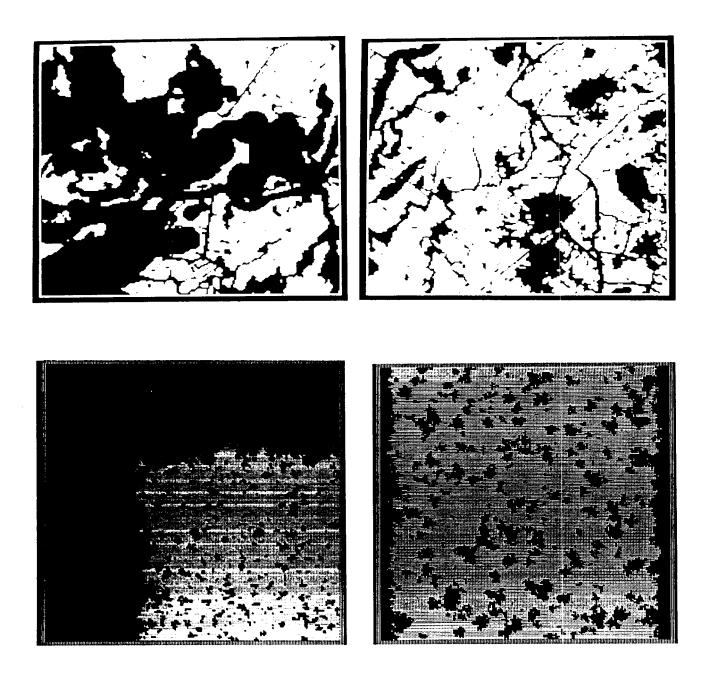


Figure E1-34. Cocodric, SW quadrangle.



W = 44 G = 540 DL = 62.97 BC = 0011 W = 213 G = 180 DL = 30.74 BC = 0101

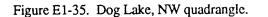
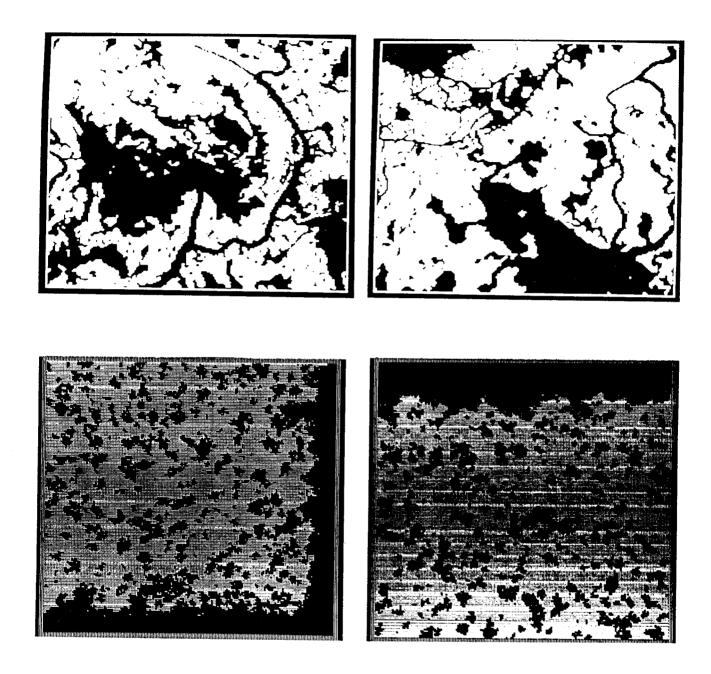
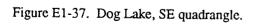


Figure E1-36. Dog Lake, NE quadrangle.



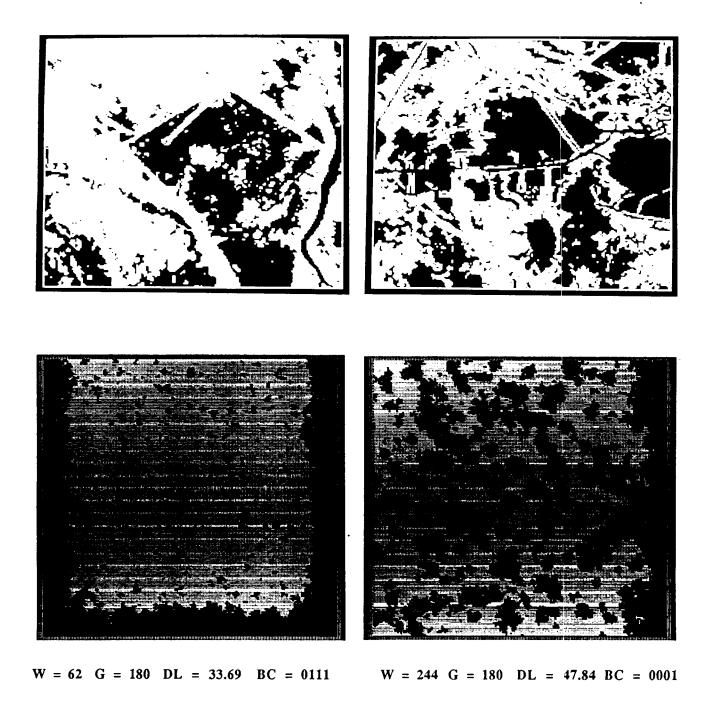
W = 120 G = 180 DL = 42.38 BC = 0011





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Figure E1-38. Dog Lake, SW quadrangle.



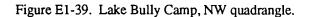
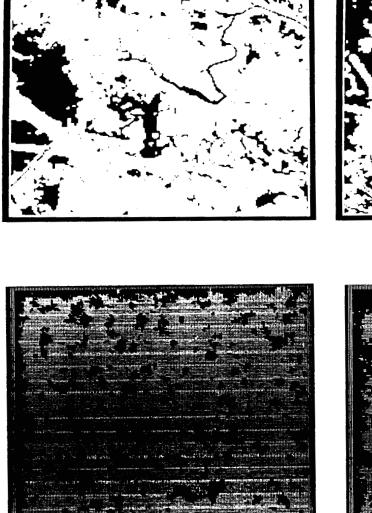
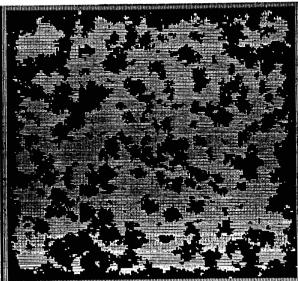


Figure E1-40. Lake Bully Camp, NE quadrangle.





W = 130 G = 60 DL = 22.21 BC = 0011

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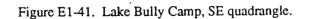


Figure E1-42. Lake Bully Camp, SW quadrangle.

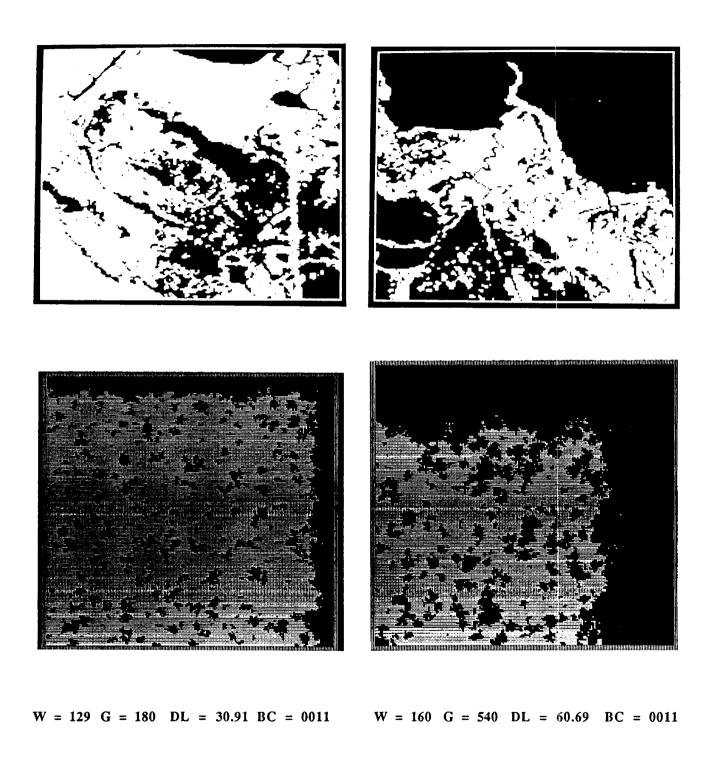
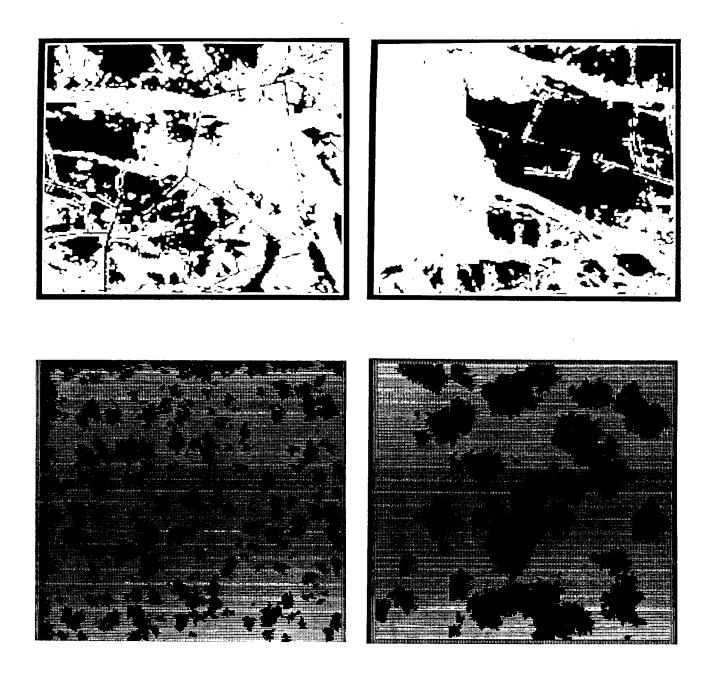


Figure E1-43. Golden Meadow Farms, NW quadrangle. Figure E1-44. Golden Meadow Farms, NE quadrangle.



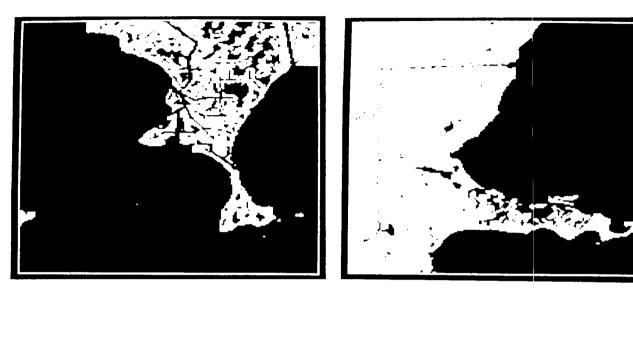
W = 308 G = 60 DL = 35.56 BC = 0011

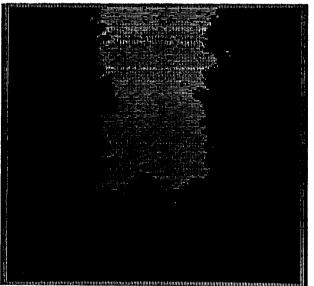
Figure E1-45. Golden Meadow Farms, SE quadrangle.

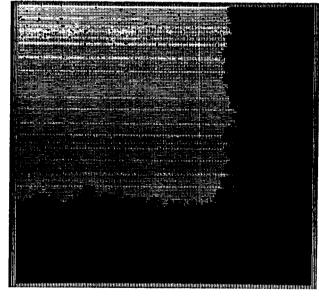
Figure E1-46. Golden Meadow Farms, SW quadrangle.

G = 0 DL = 38.95 BC = 0000

W = 3184

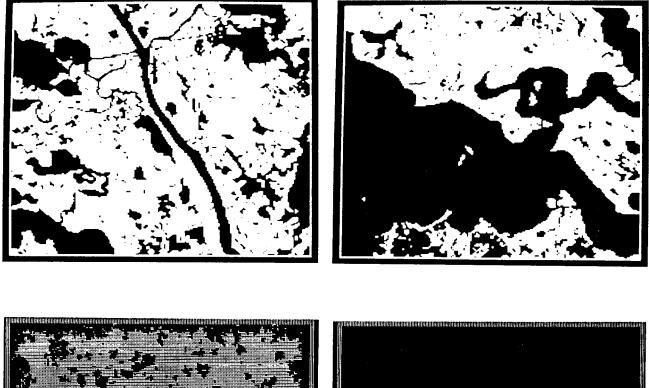


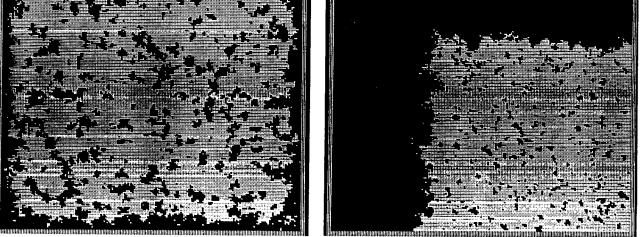




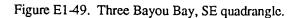


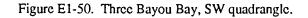




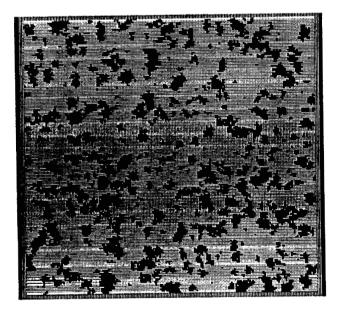


W = 100 G = 60 DL = 32.93 BC = 1111 W = 33 G = 540 DL = 58.71 BC = 0011



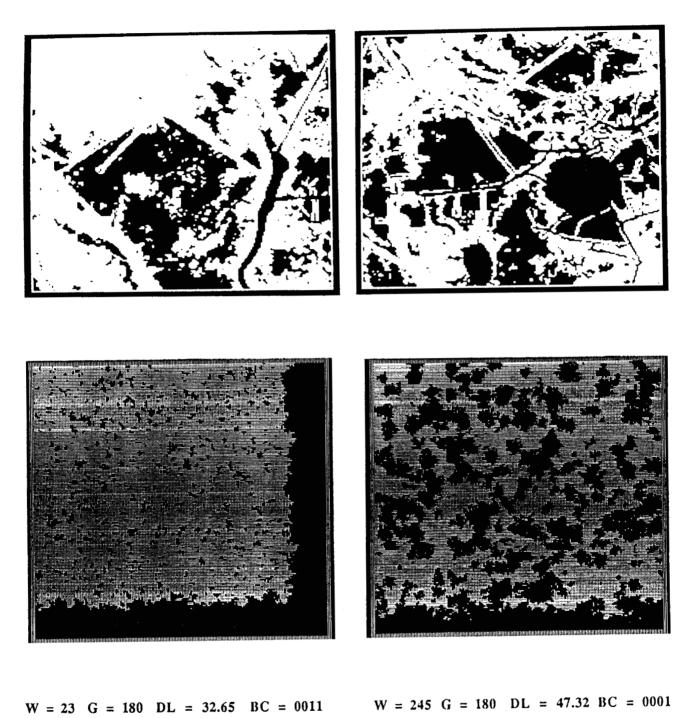






W = 116 G = 0 DL = 27.66 BC = 0011

Figure E1-51. Golden Meadow, SW quadrangle.



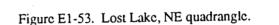


Figure E1-52. Lost Lake, NW quadrangle.

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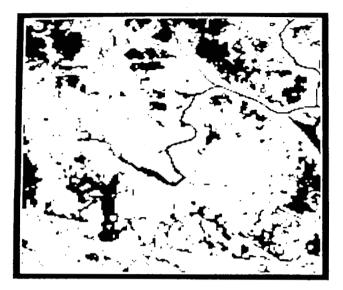
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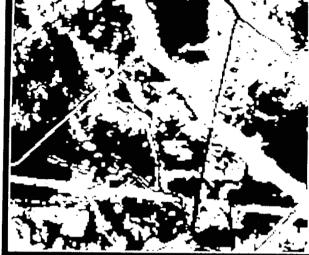
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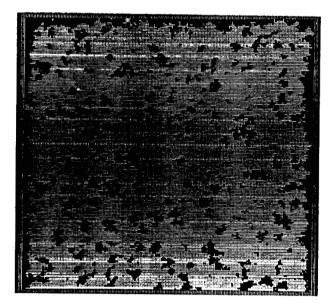
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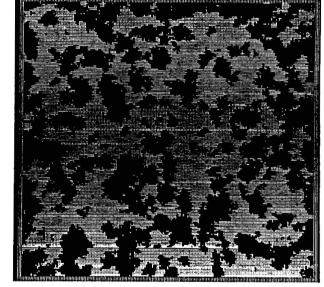
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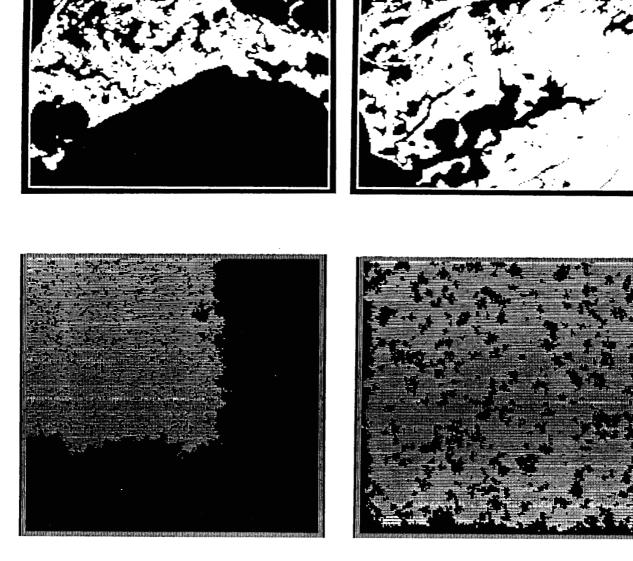


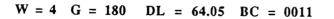
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W = 111 G = 60 DL = 21.81 BC = 0111 W = 290 G = 20 DL = 50.78 BC = 0111

Figure E1-54. Lost Lake, SE quadrangle.

Figure E1-55. Lost Lake, SW quadrangle.





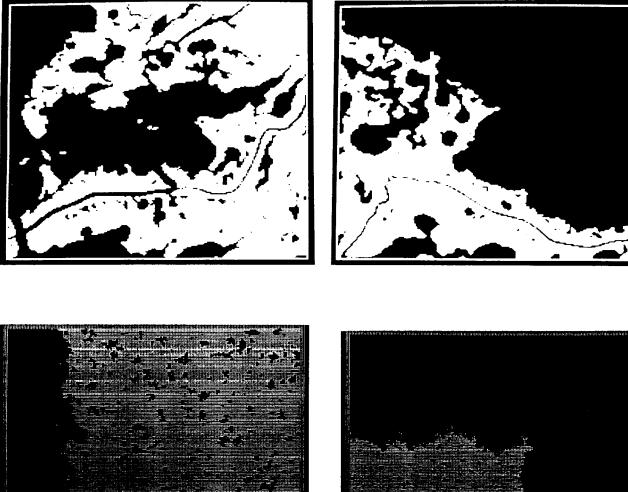


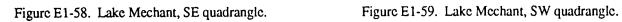


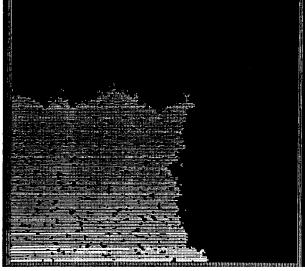
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Figure E1-57. Lake Mechant, NE quadrangle.

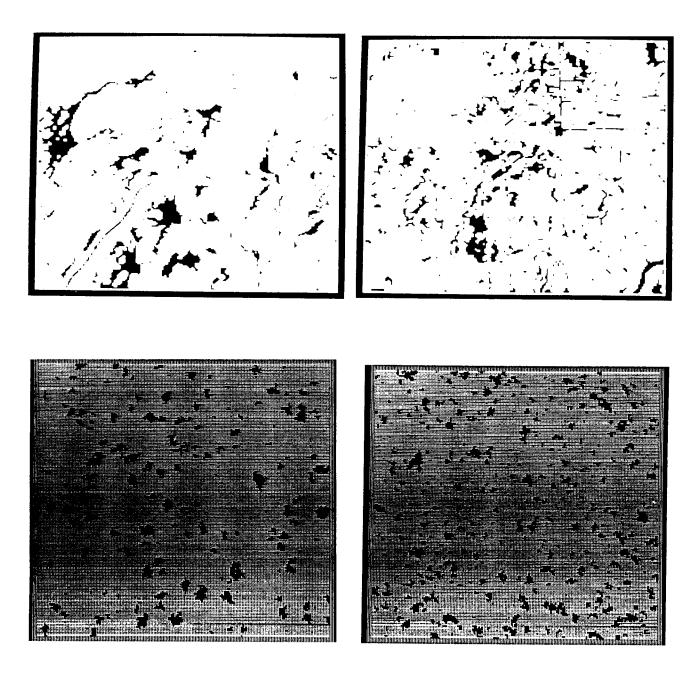




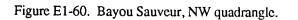








W = 325 G = 0 DL = 8.72 BC = 0000



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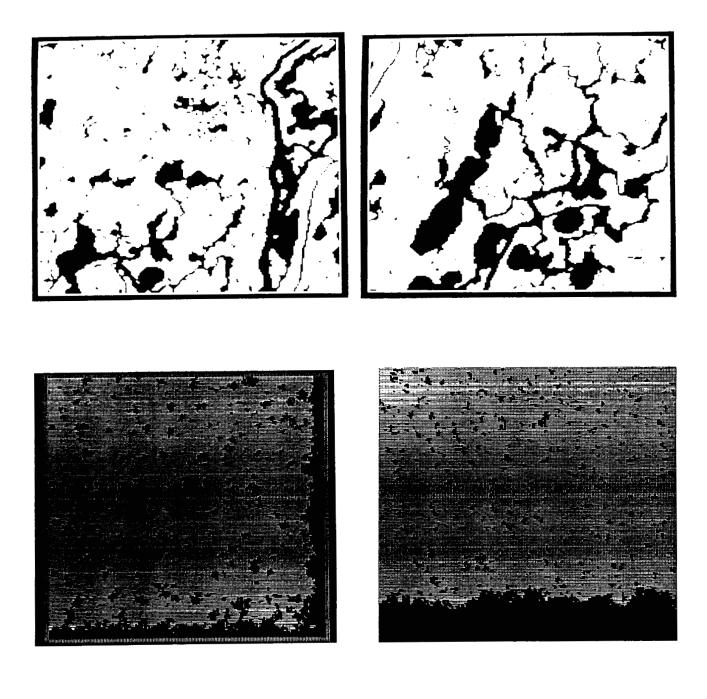
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Figure E1-61. Bayou Sauveur, NE quadrangle.

G = 0 DL = 10.90 BC = 0000

W = 93



W = 133 G = 180 DL = 22.23 BC = 0011

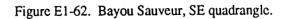
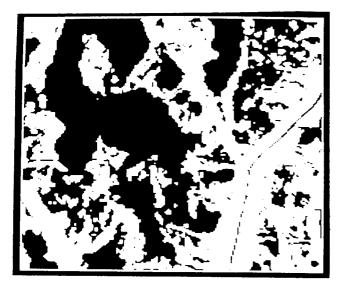


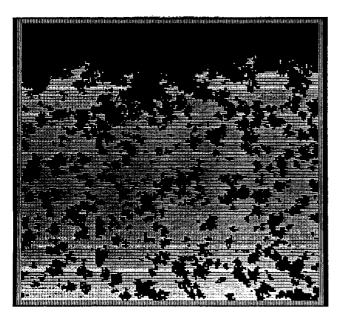
Figure E1-63. Bayou Sauveur, SW quadrangle.

W = 35 G = 540 DL = 24.31 BC = 0001



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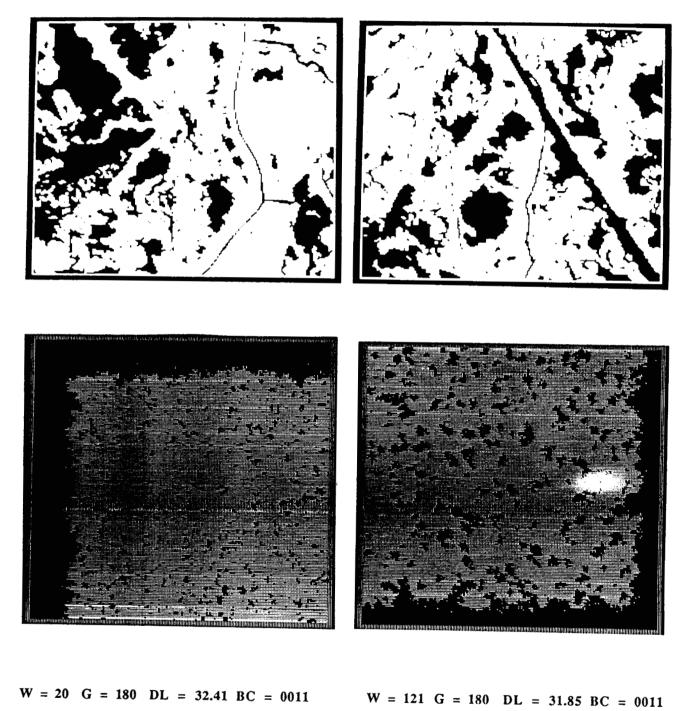
W = 116 G = 540 DL = 48.20 BC = 0001

Figure E1-64. Lake Quitman, NE quadrangle.

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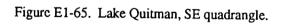
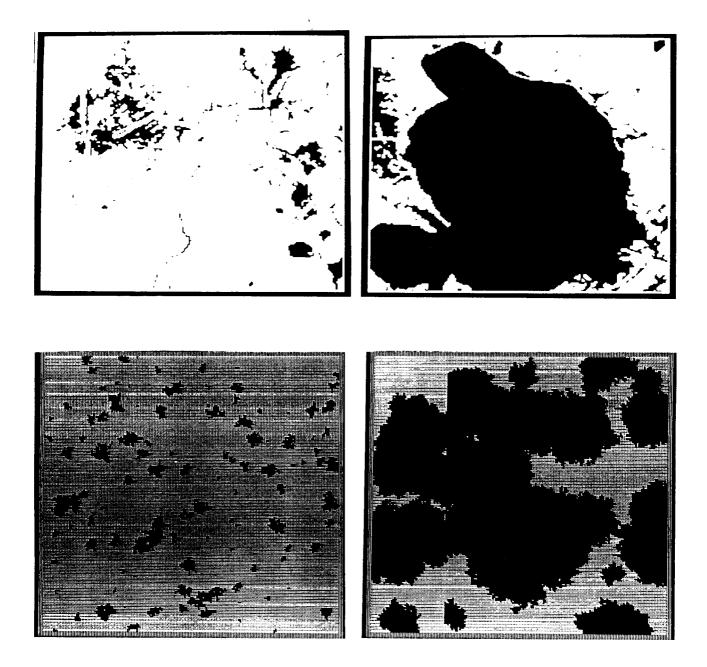


Figure E1-66. Lake Quitman, SW quadrangle.



W = 701 G = 0 DL = 9.63 BC = 0000



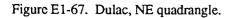
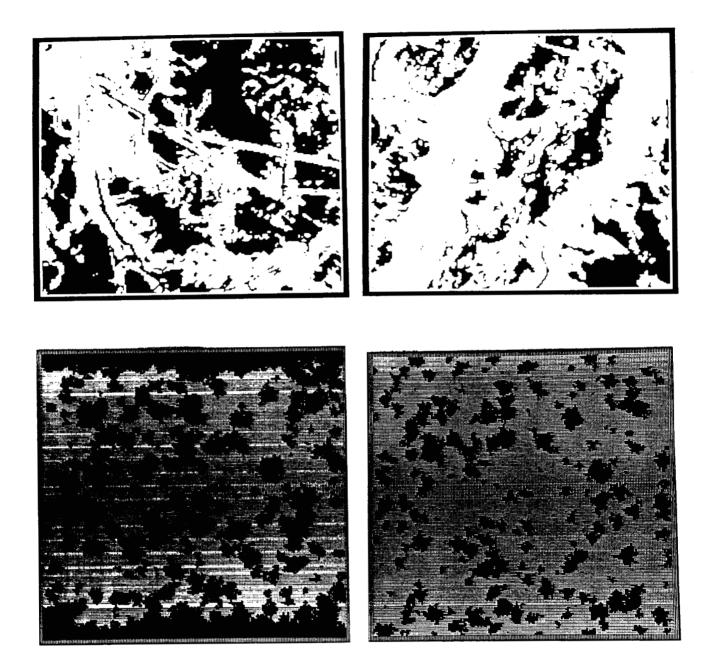


Figure E1-68. Dulac, SE quadrangle.



W = 289 G = 180 DL = 47.48 BC = 0101



Figure E1-69. Montegut, SE quadrangle.

Figure E1-70. Montegut, SW quadrangle.

ORIGINAL PAGE IS OF POOR QUALITY Figure E2. Comparisons of pond size distributions of TM scenes with the "best-fit" simulations at the same disintegration level. Water-body size is the frequency distribution of water-body classes (in pixel units) expressed as percentage of total number of water pixels. Frequency is the number of pixels in a water-body class, expressed as a percentage of the total number of water pixels. Each mirror image histogram shows one TM scene and its representative simulation. Quadrangles are grouped by age of delta lobe and salinity (as in Table D1).

Late Lafourche, salt

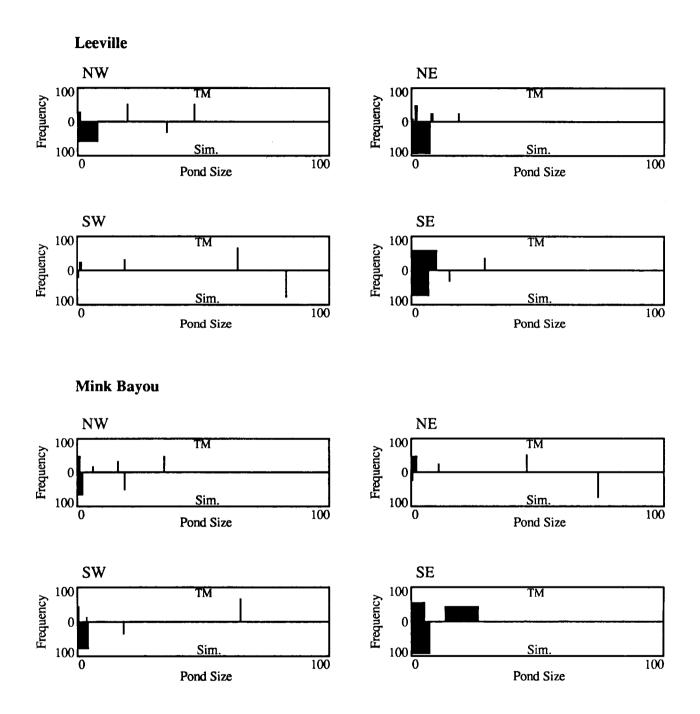


Figure E2-1. Pond size distributions for Leeville and Mink Bayou quadrangles.

Late Lafourche, salt

Caminada Pass

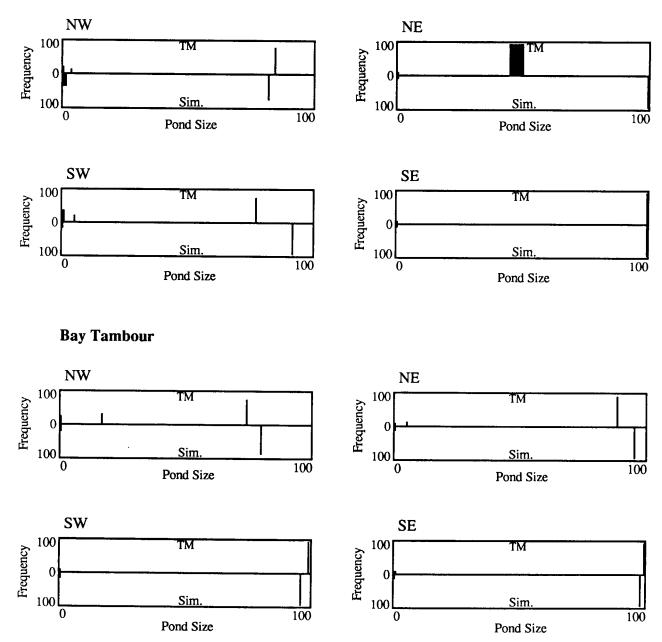


Figure E2-2. Pond size distributions for Caminada Pass and Bay Tambour quadrangles.

Late Lafourche, salt

Pelican Pass

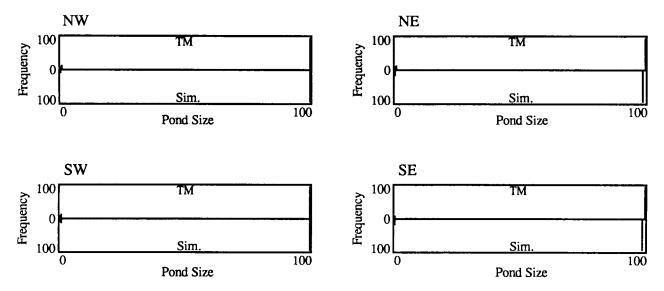
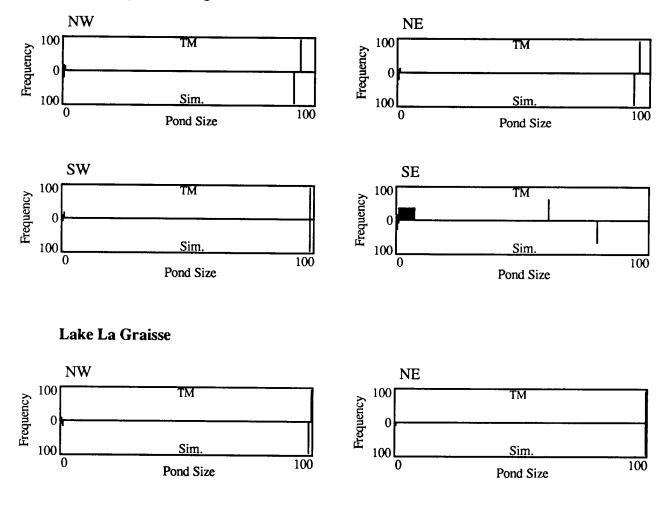


Figure E2-3. Pond size distributions for Pelican Pass quadrangle.

Early Lafourche, salt

Grand Bayou du Large





Early Lafourche, salt

Central Isles Dernieres

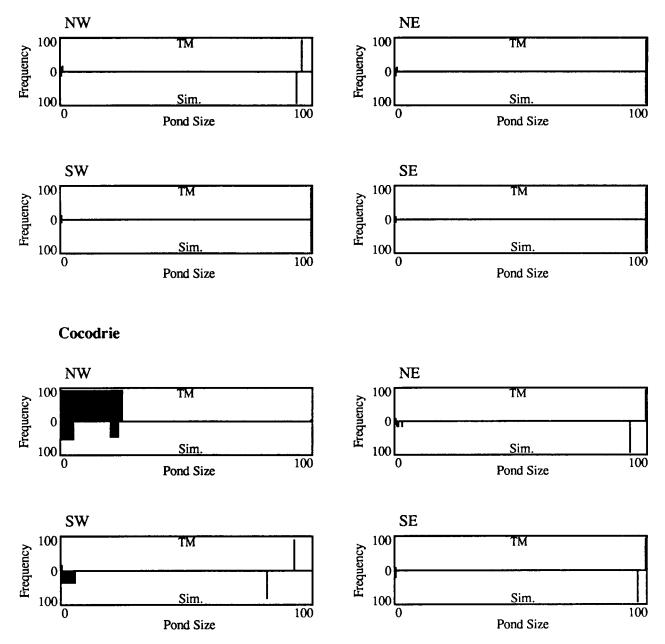
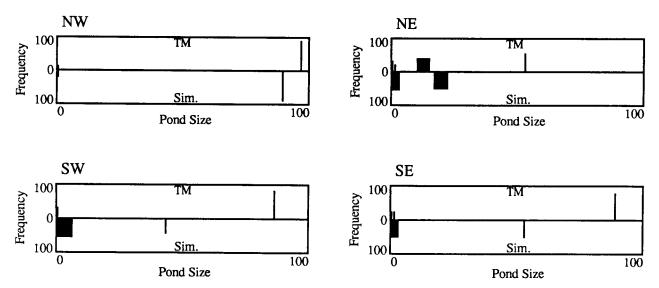
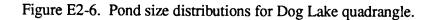


Figure E2-5. Pond size distributions for Central Isles Dernieres and Cocodrie quadrangles.

Early Lafourche, salt

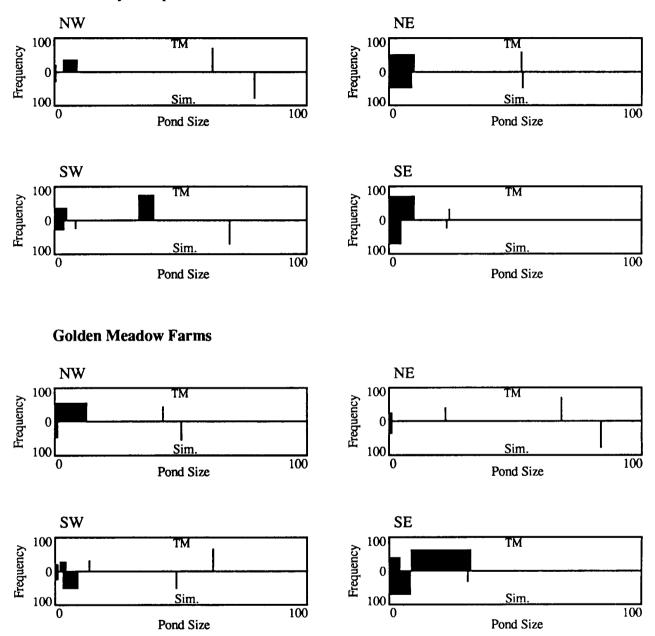
Dog Lake

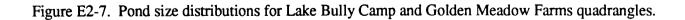




Late Lafourche, brackish

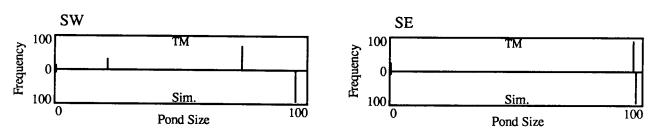
Lake Bully Camp



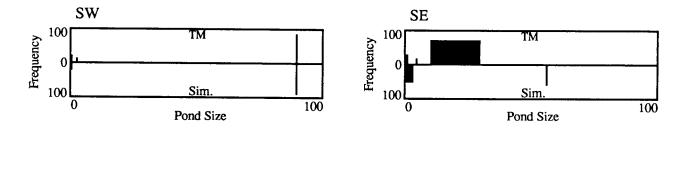


Late Lafourche, brackish

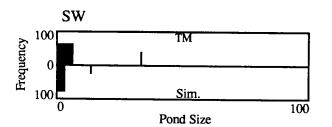
Bay L'Ours

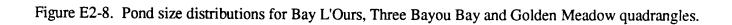


Three Bayou Bay



Golden Meadow





Early Lafourche, brackish

Lost Lake

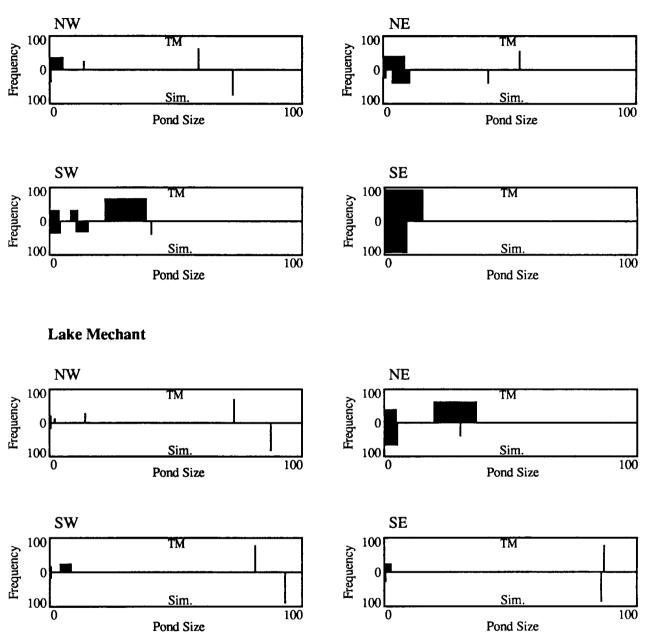
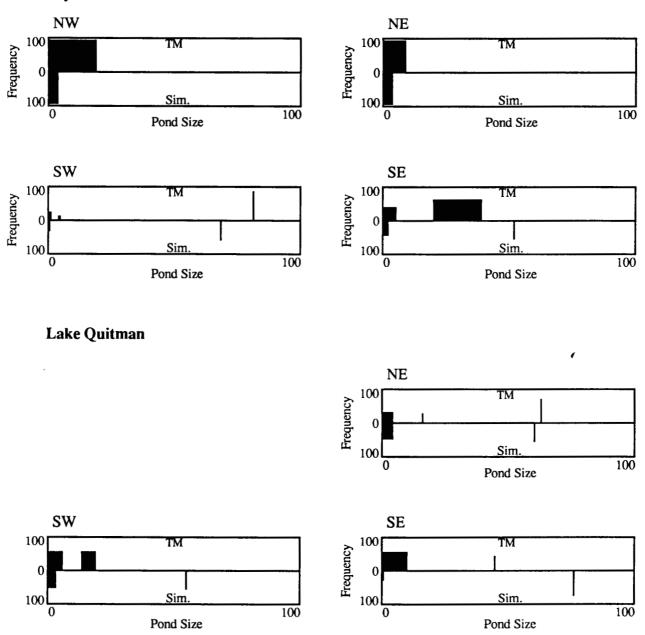


Figure E2-9. Pond size distributions for Lost Lake and Lake Mechant quadrangles.

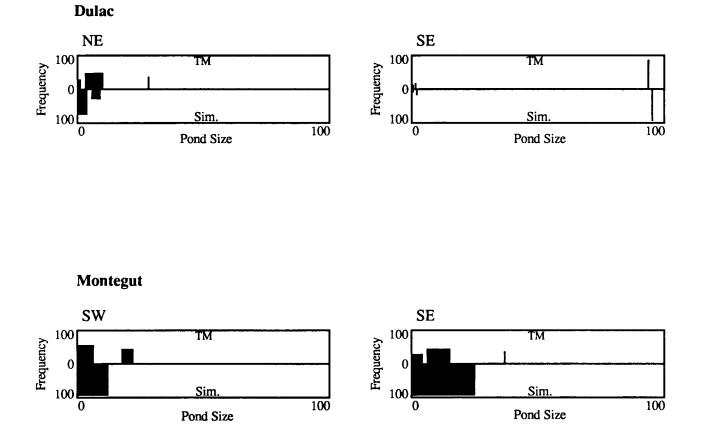
Early Lafourche, brackish

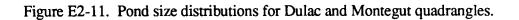
Bayou Sauveur

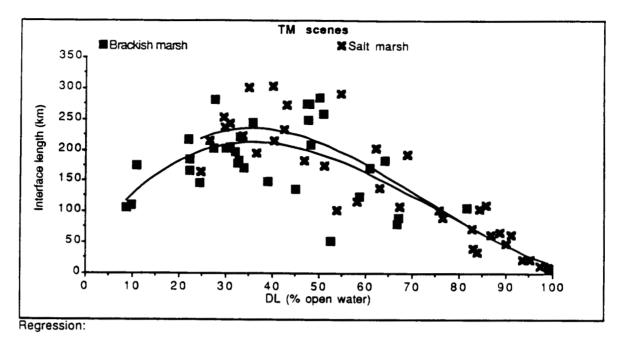




Early Lafourche, brackish







TM scenes:Brackish $y = 34.92 + 11.1x - 0.205x^2 + 0.0009x^3$; $R^2 = 0.32$ Salt $y = 23.77 + 13.26x - 0.246x^2 + 0.001x^3$; $R^2 = 0.83$

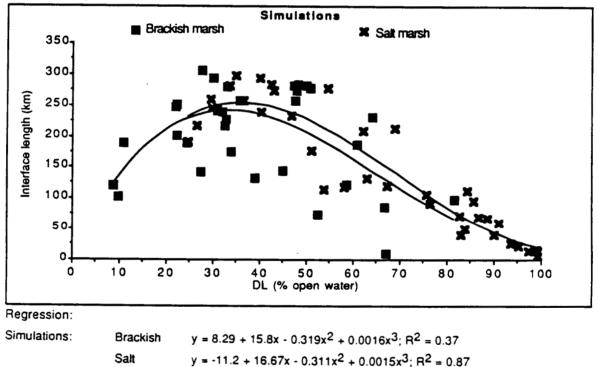


Figure E3. Comparison of interface as a function of disintegration level for the 70 TM scenes and for simulation at the same disintegration levels.

Figure E4. TM imagery interface and side adjacency statistics (number of pixels) compared with simulations at the same disintegration level. The side adjacency statistic refers to the number of sides of a water pixel that border water and may equal 0-4.

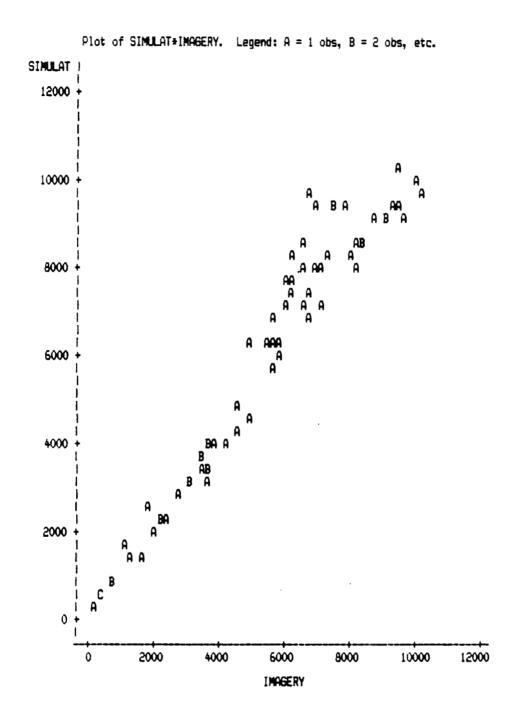
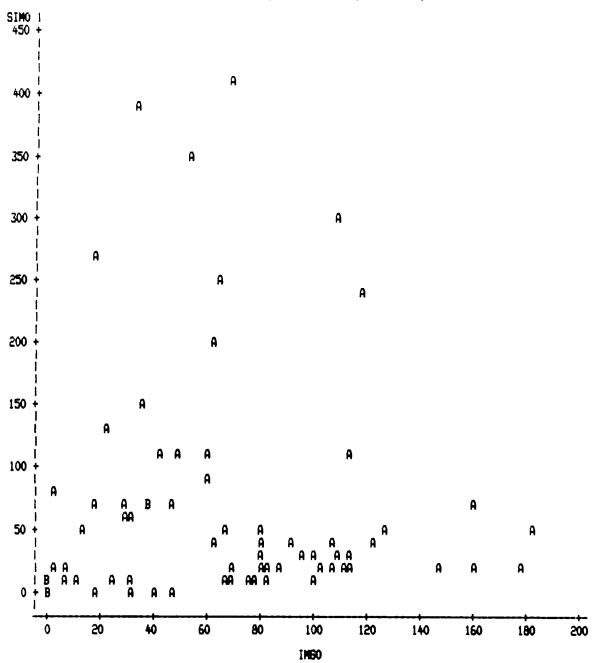
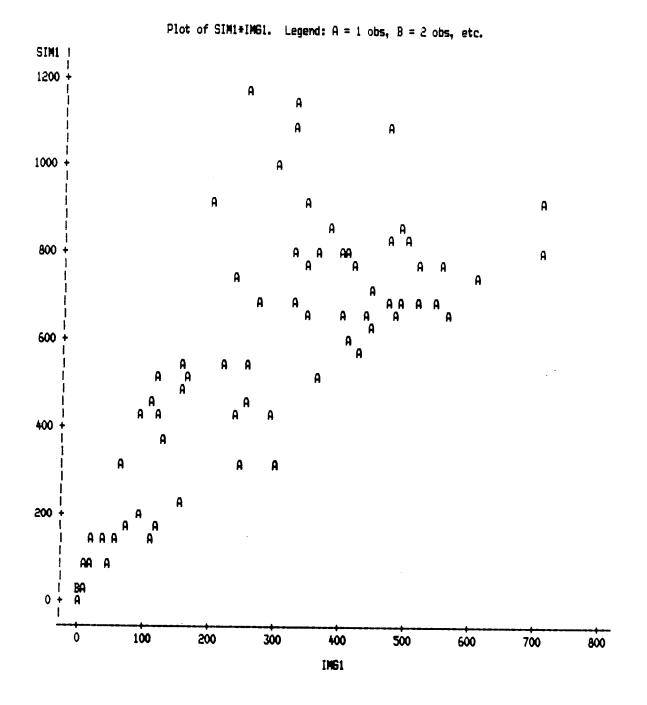


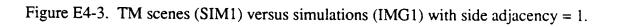
Figure E4-1. Total interface (number of pixels).

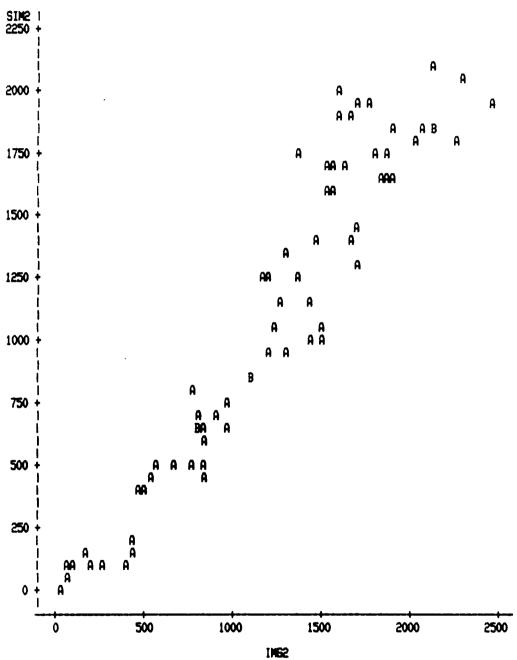


Plot of SIMO*IMGO. Legend: A = 1 obs, 8 = 2 obs, etc.

Figure E4-2. TM scenes (SIM0) versus simulations (IMG0) with side adjacency = 0.

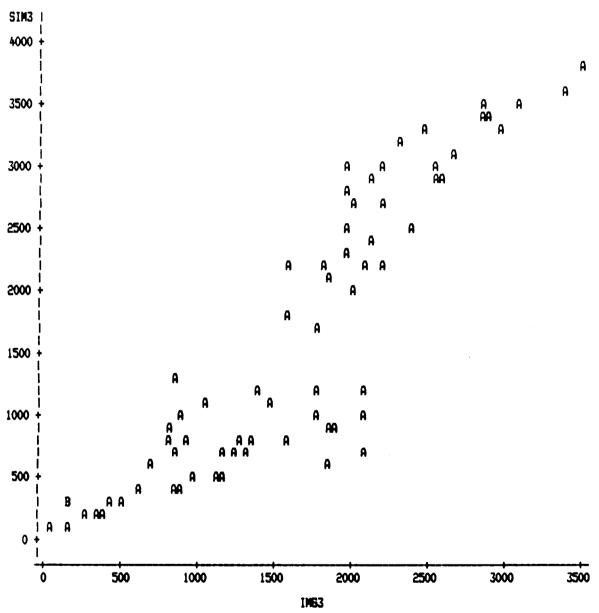






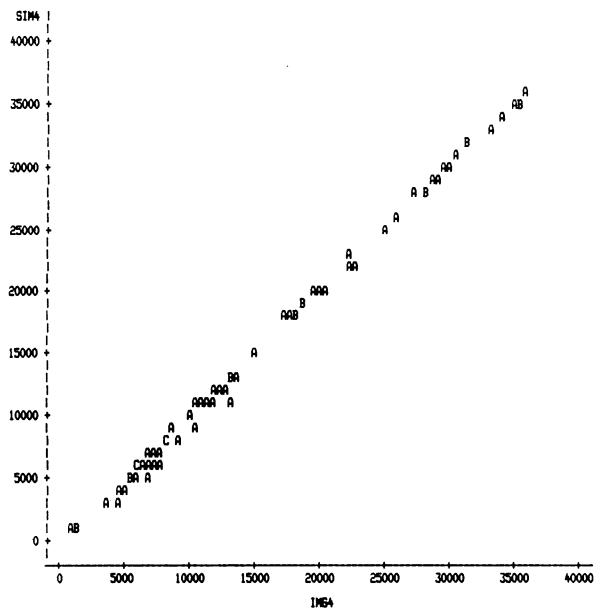
Plot of SIM2+IM62. Legend: A = 1 obs, B = 2 obs, etc.

Figure E4-4. TM scenes (SIM2) versus simulations (IMG2) with side adjacency = 2.



Plot of SIM3+IM63. Legend: A = 1 obs, B = 2 obs, etc.

Figure E4-5. TM scenes (SIM3) versus simulations (IMG3) with side adjacency = 3.



Plot of SIM4*IM64. Legend: A = 1 obs, B = 2 obs, etc.

Figure E4-6. TM scenes (SIM4) versus simulations (IMG4) with side adjacency = 4.

Appendix F

Presimulation Predictions from the Knowledge Base

Table F1. Presimulation values of the spatial-pattern indices, predicted from the knowledge base.

Presimulation Predictions from the Knowledge Base

The expert system's estimate of the model parameters W, G, and BC to "best" simulate the disintegration of a specific marsh was derived from separate estimates based on each of the spatial-pattern indices. For instance, the W-estimate is a weighted mean of W's based on matching each of several spatial-pattern indices from the imagery to those in a knowledge base. The knowledge base consists of sets of look-up tables that relate spatial-pattern index values for a given DL to W, G, and BC. The best W-G-BC estimate was determined from the knowledge base in the manner described in Appendix A. Given a W-G-BC estimate, it was possible to work backward through the selection process (relating a given W, G, and BC to spatial-pattern index values) and use the expert system to "predict" the spatial pattern indices for a given DL that would result from a simulation of marsh disintegration using that particular W-G-BC estimate. The presimulation predictions of spatial-pattern indices for the 70 cases of our study are presented in Table F1. The significance of presimulation predictions of the spatial-pattern indices is that they can potentially be used to test and fine tune W-G-BC estimates produced by the expert system. Alternative decision algorithms could be tested for their relative ability to simulate the spatial-pattern indices of a given study site. A comparison of presimulation to simulation values of the spatial pattern indices is presented in Appendix G.

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Table F-1. Pre knowledge base: 0, 1, 2, 3, and marsh border.		Pre-simulation value se: interface lengt and 4 sides adjacent . Disintegration le	values of the length, the si acent to other on level (DL,	spatial de-adja water water a	pattern cency sta pixels), rea as p	indices, tistics and numb rcentage	predicted (number of ber of water e total area	l from the water pi er pixels	e ixels with on the so shown.
Quad Qua	Quarter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Late Lafourche	e, salt	اند							
Leeville	MN	42.94	43		σ	,90	ς,	2	œ
	NE	34.93	10,044		Ч	2,036	c	,49	162
	SE	39.91	16		7	, 08	ົບ	,10	Ч
	MS	46.74	2		ω	7	723	, 72	2
Mink Bayou	MN	.0	7,284		9	1,496	2,158	,01	e
	NE	24.54	0		9	ω	696	,91	n
	SE	•	σ		S	, 82	ő	,24	n
	MS	•	σ		9	64	ß	, 73	8
Caminada Pass		62.09	6,597	46	573	1,347	1,967	ω	645
	NE		ູ		4	7	318	9,13	0
	SE				7	9	12	5,80	S
	MS	58.02	5	2	Ч	9	539	, 12	Ч
Bay Tambour	MN	51.18	5,780	134	735	980	1,057	5,32	n
	NE		Ś		σ	ω	472	6,20	9
	SE	ω.	्	8	9	2	715	0,72	Ч
	MS	84.38	4,	63	9	α	762	8,58	0
Pelican Pass	MN	98.96	336	2	32	56	91	5,53	ß
	NE	82.60	2,331	17	224	444		8,37	σ
	SE	86.68	۲,	53	c	ω	395	0,26	1
	ΜS	97.24	359	4	33	64		4,87	9

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Table F-1. (cont. 2)

Quad Quarter	ter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Early Lafourche,	le, salt	Ļt							
Grand Bayou	MN	53.83	0	136	0	e C	4	7,61	0
du Large	NE	67.26	2	σ		4	Ч	2,42	4
	SE	40.44	4	366	7	1,131	956	1,01	9
	SW	-	4	വ	ω	ω	ω	9,64	Ч
Lake La	MN	91.13	2	9		4	ω	1,70	2
Graisse	NE	99.28	4	4	2	4	2	5,69	ഹ
Central Isles	MN		3,634	81	405	641	803	25,232	
Dernieres	NE	٠	D	16		2	L.	1,63	- 1
	SE	٠	2	14		2	4	3,91	ŝ
	SW		Ч	15	9	N	0	3,30	9
Cocodrie	MN	٠	, 21	50		, 88	, 59	6,45	9
	NE	8 .	9	11	2	0	S	0,04	4
	SE	<u>ъ</u>	, 63	10	7	Ч	, 28	8,69	e
	MS	54.72	, 75	თ	ß	H	, 83	3,22	N
Dog Lake	MN	62.97	,44	80	Ч	2	606	0,35	ŝ
	NE	30.74	, 03	21	Ч	, 65	, 48	, 01	4
	SE	42.38	, 88	40	0	σ	67	, 83	5
	SW	36.40	, 14	50	ß	, 66	,27	, 35	311

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cont.
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Table

Quad Qua	Quarter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Late Lafourche, brackish	e, brac	<u>:kish</u>							
Lake Bully	MN	33.69	, 92		687	, 13	,26	, 63	
Camp	NE	47.84	,49		689	, 90	,46	, 21	e
r	SE	22.21	8,629	50	879	1,751	2,212	2,866	423
	SW	50.02	, 25		639	,87	, 50	, 64	S
Golden Meadow		30.91	66		738	, 56	, 06	, 52	4
Farms	NE		, 94		482	, 22	, 89	, 22	Ч
	SE	35.56	, 73		702	, 75	, 98	, 18	S
	MS	38.95	S		243	, 08	77,	1,04	S
Bay L'Ours	SE	81.53	, 33	4	389	7	585	, 75	σ
I	MS	52.26	ω		133	4	284	7,83	σ
Three Bayou	SE	32.93	, 73		926	S	2,811	, 63	ŝ
Bay	SW	58.71	S		519	e	641	, 20	Ч
Golden Meadow	MS	27.66	Ч		1,030	2,193	3,021	,76	c

Table F-1. (cont. 4)

Quad Quai	Quarter	DL	Interface Length	Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Border Pixels
Early Lafourche, brackish	le, br	ackish							
Lost Lake	MN	32.65	, 30	231	7	.21	9	19	<u>د</u>
	NE	47.32	9,472	18	692	1,905	3,444	11,043	336
	SE	ω	, 19	65	ß	, 63	05	2,85) co
	SW	٠	, 09	16	0	, 98	, 85	1,60	9
Lake Mechant	MN	٠	, 11	378	-	94	61	, 13	2
	NE	•	20	48	7	ဖ	, 88	4.59) LC
	SE	6.	, 69	58	Ч	90	05	3,53	00
	SW	•	, 16	141	2	ഹ	38	. 77	4
Bayou Sauveur	MN	٠	, 12	30	9	e	2	63	' m
	NE	6	, 54	108	0	, 25	,10	8	52
	SE	3.	, 39	56	ω	9	9	. 19	3
	ΜS	٠	,09	184	ß	97	79	. 87	1
Lake Quitman	NE	ω.	D	48	c	, 79	σ	02	367
	SE	8	, 22	245	ω	7	88	8,20) LO
	SW	ω	, 78	57	9	, 57	~	. 82	4
Dulac	NE	9.63	, 56	14	4	74	96	43	• •
	SE	66.80							2
Montegut	SE	47.48	8,932		9	. 80	. 23	29	α
	MS	27.16	, 65	16	654	1,543	2,473	5,200	120

Note: No presimulation prediction could be made for Dulac SE, because the model parameters for to simulate this site were determined by extrapolation rather than by interpolation from look-up tables in the knowledge base.

Appendix G

Comparison of Presimulation Predictions to Simulation Values

- Table G1. Percentage deviation of simulation values from presimulation-predicted values of the spatial-pattern indices.
- Table G2. (A) Mean percentage deviation and (B) mean absolute percentage deviation of simulation from image values and pre-simulation prediction from simulation values of spatial-pattern indices.

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Comparison of Presimulation Predictions to Simulation Values

Given the random aspect of the model, we could not expect perfect correspondence of presimulation-predicted and simulation values of the spatial pattern indices. No two simulations were alike, even when the same W-G-BC values were used. Table G1 shows the percentage deviation of simulation values from presimulation predictions (both at the same DL, which was the DL of the study site in the imagery). Average values are given at the bottom of the table. The largest average percentage deviation is in Adj-0, which we know to be the most variable index from one simulation to another.

Average percentage deviations are summarized in Table G2, which also gives average absolute percentage deviations, which provide a slightly different picture. Since deviations can be in either the positive or the negative direction, they tend to cancel each other out in averaging procedures. This is why average percentage absolute deviations are also shown.

In addition to comparisons of simulation to presimulation-predicted values, comparisons of simulation to image values are also given in Table G2. Note that percentage absolute deviations of simulation from image values are notsubstantially higher--and, in fact, are in some cases lower--than deviations of simulation from presimulation-predicted values.

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			<u> </u>	Pe	rcentage	Deviati	on ^a	
Quad ()uarter	Interfac Length	e Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Pixels
Late Lafour	<u>ccne, sa</u>							
Leeville	NW	-3,21	10.53	-4.86	-2.63	-2.72	2.38	-8.85
	NE	-1.09	-19.05	-1.96	0.59	-1.65	0.57	20.37
	SE	-3.45	60.00	-1.55	-6.53	-2.36	2.42	16.74
Minh Deven	SW	7.07	-21.73	19.67	17.93	-0.28	-1.93	5.07
Mink Bayou	NW	-0.59 4.37	8.00	$-0.90 \\ 4.87$	-2.74 9.99	2.78 4.89	-0.20 -2.37	0.00 - -1.16
	NE SE	-3.23	-5.58 -28.00	4.87	-4.44	4.09	2.94	-17.91
	SW	2.13	-30.77	4.96	-0.31	-0.52	-2.58	0.70
Caminada Pa		5.94	-28.26	16.40	3.64	3.76	-1.26	5.27
Califinada Po	NE	-5.37	7.04	-6.43	-5.59	-11.95	0.20	0.28 1
	SE	100.00	33.33	57.14	250.00	341.67	-0.11	-2.27
	SW	5.68	-15.87	4.87	16.40	7.98	-0.62	-2.52
Bay Tambour		1.80	-17.16	1.63	3.37	9.08	-0.77	0.95
	NE	1.06	-10.57	10.18	-2.05	-0.21	-0.04	0.71
	SE	11.63	-12.50	-1.80	19.63	16.22	-0.58	-0.28
	SW	10.45	-24.73	23.08	14.36	4.46	-0.61	0.00
Pelican Pas	s NW	58.93	-85.71	6.25	46.43	186.31	-0.50	-1.19
	NE	2.75	-35.29	-8.04	8.56	11.55	-0.33	1.30
	SE	10.61	-15.09	26.36	8.57	4.05	-0.35	1.15
	SW	23.12	75.00	21.21	25.00	25.23	-0.14	0.00
Early Lafou	urche, sa	alt						1
Grand Bayou	1 NW	2.62	-16.18	1.39	15.30	-0.73	-0.36	0.79
du Large	NE	8.82	-23.96	13.38	9.36	19.02	-0.93	-0.55
-	SE	4.30	-18.58	10.92	12.91	-1.99	-1.53	1.51
	SW	2.01	5.98	-5.68	8.54	-12.22	0.01	0.81
Lake La	NW	11.91	-66.67	5.71	21.47	9.52	-0.43	-0.28
Graisse	NE	100.00	-75.00	4.76	39.02	313.70	-0.66	-0.40
Central Is	les NW	-2.37	-28.40	-5.68	4.37	3.61	-0.02	0.00
Dernieres		-9.81	18.75	5.15	-19.42	-13.46	0.29	1.68
	SE	-0.42	-21.43	35.19	-4.00	-12.86	0.07	0.14
	SW	18.43	13.33	15.63	33.61	10.24	-0.20	0.00
Cocodrie	NW	1.76	-20.00	-2.53	3.51	5.75	-2.65	-1.28
	NE	-2.09	-63.64	1.64	-0.57	-4.15	0.72	-1.13
	SE	-10.84	0.00	-16.73	-7.96	-9.55	0.94	-5.84
-	SW	-4.85	-22.22	-9.10	-4.96	-1.91	2.03	-6.45
Dog Lake	NW	-0.97	-28.75	-5.88	5.72	5.53	-0.18	-0.56
	NE	1.17	19.05	0.00	0.79	2.42	-1.35	2.03
	SE	6.20	-35.00	-0.25	7.63	13.04	-4.58	-3.37
	SW	5.88	-10.00	6.22	2.76	12.02	-4.39	3.22

Table G1. Percentage deviation of simulation values from presimulationpredicted values of the spatial-pattern indices. Table G1. (cont.)

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				P	ercentag	e Deviat	ion ^a	
Quad Qua	rter	Interfac Length	ce Adj-0	Adj-1	Adj-2	Adj-3	Adj-4	Pixels
Late Lafourch	e, bra	ackish				·		
Lake Bully	NW	-1.50	-5.33	-1.31	-1.05	-1.66	0.52	0.98
Camp	NE	-4.12	-50.00	-3.92	-5.56	-1.88	1.73	5.62
	SE	-3.23	-12.00	-8.42	-1.09	0.54	2.97	1.89
	SW	1.59	18.18	5.79	-1.71	2.88	-0.92	0.53
Golden Meadow		4.97	-26.79	3.12	8.33	6.01	-3.80	-1.81
Farms	NE	5.02	-23.08	6.22	2.61	9.28	-1.25	0.19
	SE	-1.96	18.75	-2.71	-0.23	-3.72	1.84	1.09
Dave T ! Ourse	SW	-10.58	-93.62	31.28	-21.38	-2.48	2.81	-42.14
Bay L'Ours	SE	-0.75	-28.19	13.88	8.39	-14.02	0.14	-0.57
Three Bayou	S₩ SE	-0.21 -4.39	3.76 -29.17	3.76 -10.26	-15.75 -3.63	-5.28	0.12	0.00
Bay	SW	3.84	-25.44	2.70	-3.63	2.70 3.59	1.90 -0.51	0.66 -3.47
Golden Meadow		-5.77	-23.44	-11.75	-5.24	0.27	6.54	-5.84
Early_Lafourc	he, bi	rackish						
Lost Lake	NW	3.62	-13.85	11.13	1.15	5.00	-1.64	0.00
	NE	-0.90	-11.11	-2.89	-2.21	1.97	-0.08	3.27
	SE	0.61	-16.92	-3.41	2.88	4.98	-3.61	0.17
	SW	-8.32	-37.50	-12.18	-6.41	-7.92	4.58	-0.89
Lake Mechant	NW	8.21	8.99	12.46	3.50	-2.94	-0.98	4.92
	NE	-4.02	-18.75	-12.28	-3.01	3.95	1.52	2.88
	SE	3.13	15.52	1.94	4.88	2.93	-0.52	-3.53
Paulou Caunaur	SW	1.93	-4.97	3.29	4.41	3.41	-0.08	-3.14
Bayou Sauveur	NW NE	-3.15 -3.51	-20.00 -36.11	-6.52 -3.60	-2.03 -1.03	2.28 2.71	3.12	28.13
	SE	5.41	-19.64	-0.29	5.47	16.98	5.00 -7.46	46.15 -0.71
	SW	3.76	-19.57	5.99	9.21	4.80	-2.52	3.66
Lake Quitman	NE	6.92	-50.00	4.60	4.91	14.94	-4.57	11.17
2	SE	0.53	2.86	0.20	-0.43	1.47	-0.22	0.66
	SW	2.81	-36.84	4.87	0.00	10.08	-3.27	0.67
Dulac	NE SE	-3.45	-28.57	-9.59	-4.58	5.89	1.53	-7.50
Montegut	SE	-4.27	35.71	1.21	-7.00	-4.69	2.47	-1.87
	SW	-2.86	-12.50	-11.62	2.20	-2.26	1.98	2.50
Mean		5.07	-15.40	3.39	7.56	14.52	-0.14	0.76

Table G2. (A) Mean percentage deviation and (B) mean absolute percentage deviation of simulation from image values and presimulation predictions from simulation values of spatial-pattern indices.

A. Percentage deviation

	Me	an	Stand	. Dev.
Spatial Pattern Index	Simulation from Image	Pre-sim-pred from Simulation	Simulation from Image	Pre-sim-pred from Simulation
Interface	12.28	18.80	5.07	18.91
Adj-0	105.15	393.28	-15.40	28.62
Adj-1	172.34	191.04	3.39	12.21
Adj-2	-14.28	19.28	7.56	31.80
Adj-3	-3.50	39.44	14.52	59.38
Adj-4 Border	-4.44	8.67	-0.14	2.32
water pix	29.10	44.09	0.76	9.62

B. Percentage absolute deviation

	Me	an	Stand	. Dev.
Spatial Pattern Index	Simulation from Image	Pre-sim-pred from Simulation	Simulation from Image	Pre-sim-pred from Simulation
Interface	14.54	8.29	17.08	17.72
Adj-0	189.14	25.68	359.97	19.76
Adj-1	172.34	8.46	191.04	9.38
Adj-2	19.21	11.72	14.30	30.50
Adj-3	32.43	18.10	22.37	58.38
Adj-4	5.59	1.63	7.96	1.64
Border				
water piz	c 38.13	4.40	36.44	8.58

Appendix H

Levels of Disintegration by USGS Topographic Map in 1956 and 1978 Data Compiled by Leibowitz (LSU) from Maps by Wicker (1980) with Annual Trend

Table H1. Levels of disintegration in 1956 and 1978 for the areas covered by each U.S. Geological Survey topographic map corresponding to our study sites, and annual trend in DL.

Table H1. Levels of disintegration (DL, water area as percentage total area of site) in 1956 and 1978 for the areas covered by each U.S. Geological Survey topographic map corresponding to our study sites^a, and annual trend in DL.

		Year	_
Quadrangle	1956	1978	Annual Trend
Late Lafourche, s	alt		
Leeville	20.348	34.866	-0.659
Mink Bayou	16.195	32.606	-0.745
Caminada Pass	74.764	78.267	-0.159
Bay Tambour	68.500	74.506	-0.273
Pelican Pass	88.324	91.086	-0.125
Early Lafourche,	<u>salt</u>		
Grand Bayou			
du Large	62.358	65.693	-0.151
Lake LaGraisse	93.909	96.826	-0.132
Central Isles			
Dernieres	84.760	87.623	-0.130
Cocodrie	50.914	63.774	-0.584
Dog Lake	38.994	52.395	-0.609
Late Lafourche, b	rackish		
Lake Bully Camp	4.294	26.711	-1.018
Golden Meadow			
Farms	16.421	41.627	-1.145
Bay L'Ours	40.193	49.211	-0.409
Three Bayou Bay	24.009	40.360	-0.743
Golden Meadow	7.502	16.180	-0.394
Early Lafourche,	<u>brackish</u>		
Lost Lake	29.838	35.709	-0.266
Lake Mechant	31.240	54.635	-1.063
Bayou Sauveur	10.813	25.621	-0.673
Lake Quitman	31.498	44.509	-0.591
Dulac	13.314	32.425	-0.868
Montegut	9.038	33.700	-1.121

Note: Mean annual trend, late Lafourche lobe = -0.567% Mean annual trend, early Lafourche lobe = -0.563%

^a Calculated by the authors from original data compiled by Liebowitz (LSU, private communication, 1988) from maps by Wicker (1980).

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Appendix I

Shrimp Catch, 1960-1987, in Barataria and Terrebonne-Timbalier Bays

Table I1. Shrimp catch in Barataria Bay and Terrebonne-Timbalier bays, annual rainfall, and number of hours between April 9 and 30 when temperatures were below 20°C, 1960-1987.

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Table I1. Shrimp catch (pounds) in Barataria Bay (area = 57,709 acres) and Terrebonne-Timbalier bays (130,101 acres), annual rainfall (inches), and number of hours between April 9 and April 30 when temperatures were below 20°C, by year, from 1960 through 1987.

	Ca	tch	Rair	nfall	
Year	Barataria	Terr-Timb	Barataria		Hours
1960	3,145,172	4,415,118	56.2	46.4	32
1961	1,331,299	2,887,287	73.4	78.6	148
1962	1,407,457	2,507,703	40.4	37.0	37
1963	3,043,799	5,476,282	69.3	51.4	19
1964	1,386,736	2,565,840	72.2	62.3	100
1965	3,625,101	4,283,412	60.6	59.0	0
1966	3,356,306	3,697,933	81.9	72.6	34
1967	4,947,377	6,376,487	65.3	61.7	0
1968	5,538,431	6,759,208	50.8	52.0	0
1969	5,343,343	7,030,628	56.9	58.0	4
1970	6,101,020	6,549,845	55.7	72.1	27
1971	6,243,131	9,276,749	59.4	64.5	11
1972	6,015,595	6,900,478	62.2	70.9	0
1973	3,703,811	5,848,112	77.9	73.4	137
1974	5,159,804	3,935,266	56.2	62.7	18
1975	2,503,492	5,311,733	71.4	72.6	95
1976	7,070,085	8,781,531	47.7	58.4	0
1977	5,480,952	11,011,931	63.9	79.6	9
1978	4,084,009	8,985,448	65.1	61.1	49
1979	3,637,205	6,377,519	72.8	75.3	0
1980	3,340,586	3,638,863	76.8	74.6	92
1981	5,185,351	8,845,254	50.7	54.6	0
1982	5,564,216	8,248,135	67.0	67.7	80
1983	5,488,142	5,937,343	61.8	63.5	205
1984	6,028,856	6,968,363	61.8	64.5	86
1985	3,965,987	4,217,895	68.4	70.1	61
1986	8,532,259	4,736,575	49.4	59.7	7
1987	6,463,704	5,046,182	70.1	65.9	103

Note: Barataria Bay is associated with the late Lafourche lobe, and Terrebonne and Timbalier bays are associated with the early Lafourche lobe. The rainfall station used for the Barataria area was Houma, and the station used for the Timbalier-Terrebonne area was primarily Golden Meadow, supplemented by Galiano and Paraday. Shrimp catch data were compiled from catch-by-area data of the National Marine Fisheries Service, Miami, Florida (G. Davenport, private communication, 1988). Rainfall data were provided by Robert Muller of Louisiana State University (private communication, 1988). Hours of temperatures below 20°C were obtained from Barrett and Gillespie (1975) and B. Barrett of the Louisiana Department of Wildlife and Fisheries (private communication, 1988).

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