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A Comparison of SPOT Simulator Data with Landsat MSS Imagery for Delineating Water Masses in Delaware Bay, Broadkill River, and Adjacent wetlands*

RESEARCH CONDUCTED in tidal wetlands and adja-
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INTRODUCTION only to delineate and monitor these areas from year to year, but also to make assessments based upon

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productivity. Evidently, nutrient flux from salt smarsh and estuarine research. However, the use of

ABSIIL~CT: *The raclioinetric und sputicll clualities* **of** *SPOT siinrclator cind Landsat-3 MSS data are compared as to their ability to distinguish different water masses* within Delaware Bay and adjacent wetland areas. The *SPOT simulator data contain* a greater range of gray level values for all water areas than do the Landsat MSS data. The greater spatial resolution of the *SPOT* simulator data provides information about small-scale hydrodynamics not available on the Landsat MSS data. Both types *of data show a pl~~n~e of spectmll!l riniqtle water flowing froin Roosecelt Inlet into Delaware Bay. The plume is most visible in SPOT simulator band 1 (500-590 nm) and Landsat am band* **4** (500-600 *nnz). In both bands, the pltliize appears dark relative to the surrounding Delaware Bay water. Recent hydrographic surveys characterize the plume as an ebb tidal feature with high concentrations of dissolved* and particulate organic matter believed to originate from the adjacent Canary *Creek Marsh and Great Marsh*. *spor simulator data are found to delineate water Creek Marsh and Great Marsh. SPOT simzclator data are found to delineate water rnasses with a high clegree of separation. Radionaetrically degraded SPOT data pro*duce similar results. Landsat-3 MSS data, although useful for delineating water masses, do not produce good separation because of sensor noise.

tuarine and coastal food chains. Because tidal wet-
lands tend to occupy areas ideal for such diverse human activities as industrial development, recrea-
tion, and even waste disposal, there is a need not
delineating salt marsh boundaries, and tion, and even waste disposal, there is a need not

* **Preselited at the 1984 SPOT Symposium, Scottsdale. the range of gray levels representing water to 10** Arizona, 20-23 May 1984.

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marshes can constitute a significant input to es-
tuarine and coastal food chains. Because tidal wet-
applications for two reasons:

- the spatial resolution (1.1 acres) is not suitable for detecting small-scale water features or accurately
- the radiometric sensitivity is low, thus compressing the range of grav levels representing water to 10

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When the first **SPOT** satellite is launched, it will carry into orbit a high resolution scanner that will image the Earth in three bands, two visible and one infrared, that are spectrally similar to the Landsat **MSS.** However, the spatial resolution and radiometric sensitivity of the **SPOT** scanner will be much higher than the Landsat **MSS.** The purpose of this work is to compare **SPOT** simulator data collected over the south portion of Delaware Bay with Landsat-3 imagery of the same area. Emphasis is placed upon identifying estuarine water masses. The comparison will be based upon the ability to distinguish between river and salt marsh water with high concentrations of detritus and estuarine water with a lower organic content.

BACKGROUND

Numerous researchers have used Landsat **MSS** imagery to deduce and monitor estuarine water quality. Typically, regression techniques are used to relate radiance recorded in the satellite data to water measurements of turbidity, suspended sedi-
ment load, chlorophyll-a concentration, and diffuse and beam attenuation. In Delaware Bay, Klemas *et* al. (1974) digitally processed ten Landsat-1 MSS images representing different portions of the tidal phase. Coincident with each overpass, water samples were collected and analyzed for suspended sediment concentration. Regression curves were developed and used to map suspended sediment concentration over the entire bay. By comparing sediment distribution with tidal phase, Klemas et al. were able to infer surface current circulation.

Ackleson (1981) expressed Landsat **MSS** count values in band **4** (500-600 nm) and band 5 (600-700 nm) in terms of the ratio of irradiance backscatter to diffuse attenuation within the surface waters of Delaware Bay and the coast of New Jersey. It was found that, within the Landsat imagery, at least two optically distinct water masses could be identified: one with high concentrations of dissolved and particulate organics, and one lower in organic concentration, although not necessarily lower in total seston. The organic-rich water was found to originate within wetland areas and enter coastal and estuarine waters as a result of tidal flushing.

Klemas et al. (1981) investigated the use of Landsat **MSS** to identify organic and inorganic material within the near-surface waters of Delaware Bay and along the coast of New Jersey. Although the results are promising, the Landsat data proves to be of insufficient resolution to detect the smallscale variations common in Delaware Bay.

STUDY SITE DESCRIPTION

Roosevelt Inlet is located on the southern shore of Delaware Bay just north of Lewes, Delaware. As seen in Figure 1, the Broadkill River, Canary

FIG. 1. Canary Creek study site. The dashed lines delineate the area imaged during the SPOT simulation over**flight.**

Creek, and'the Lewes-Rehoboth Canal converge at the Inlet. Although currents through the Inlet are tidal-dominated, there is a net flux of water from the Inlet to Delaware Bay as a result of fresh water input to the Broadkill River and, to a lesser extent, Canary Creek. The mean tidal range at Roosevelt Inlet is 1.3 m with a spring range of 1.6 **m.**

DeWitt and Daiber (1973) studied the hydrology of the Broadkill River and estimated that 85 percent of its 17.4-km length was bordered by salt and fresh water marshes. It is apparent that these tidal wetlands contribute significant amounts of organic materials to the Broadkill River and ultimately to Delaware Bay via Roosevelt Inlet. Roman (1981), monitoring detrital flux at the mouth of Canary Creek, reported that 10 percent of the aboveground biomass within the Spartina-dominated Canary Creek is exported annually to Delaware Bay in the form of particulate carbon. Another 31 percent enters Delaware Bay in the form of dissolved organic carbon. These detrital fluxes are manifested during ebb tide as a plume of spectrally unique water emerging from

METHODOLOGY

SPOT simulator and Landsat-3 **MSS** data are digitally processed and compared based upon the ability to distinguish different water masses in Delaware Bay. The **SPOT** simulator data was collected over Canary Creek Marsh, Delaware, on 7 July 1983. The flight line, shown in Figure 1, was 12.6 km in length and centered 2.3-km northwest of Roosevelt Inlet $(N38°48'29''$, $W75°10'31'')$. The data were collected from an altitude of 21,000 feet and the cross-track field of view of the scanner was 42.96". This resulted in a swath width of 4.7 km. The sky was free of clouds and low in atmospheric haze. At the time of the overflight, the tide at Roosevelt Inlet was beginning to flood; low slack tide occurred 1.7 hours prior to the overflight.

The Landsat-3 image was generated on 3 July 1979 and selected for comparison because it represents seasonal, atmospheric, and tidal conditions at Canary Creek that are similar to those represented in the **SPOT** data. The sky was free of clouds and low in atmospheric haze. The tidal phase at Roosevelt Inlet was flood, 1.2 hours past low slack tide.

The approach is to use an unsupervised classifier

to group the data into a limited number of water classes based upon spectral and spatial characteristics. Because the **SPOT** simulator and Landsat-3 data differ in spatial resolution and radiometric sensitivity, it is important to identify the contributions that each makes to differences in classification. This is accomplished by degrading the **SPOT** simulator data spatially and radiometrically to approximate Landsat-3 **MSS** data prior to applying the classifier. All data processing is performed using an ERDAS 400 computer system together with software either supplied by ERDAS, Inc. or developed at the University of Delaware.

Data processing applied to the **SPOT** simulator and Landsat-3 **MSS** data is shown schematically in Figures 2A and **2B,** respectively. The analysis is divided into two geographical regions based upon the degree of spatial degradation applied to the **SPOT** data. Roosevelt Inlet refers to a 235 pixel by 240 line subregion of the **SPOT** image centered at Roosevelt Inlet. The ground resolution is 20 metres. Canary Creek refers to the entire **SPOT** image after having been spatially degraded to 60 metres. This is accomplished by averaging blocks of 9 pixels, **3** pixels per side. Geographically similar regions are subsetted from the Landsat image and geometrically corrected to approximate 60-metre resolution.

FIG. 2. Diagrams showing the progression of machine processing applied to the spot simulator (A) and **Landsat-3** MSS **(B) data.**

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The **SPOT** simulator data are degraded radiometrically to approximate the Landsat-3 **MSS** data by regressing geometrically similar pixels within the two data types. Because the Landsat data are not atmospherically corrected, the regression analysis also simulated the same atmospheric effects within the data. **SPOT** band 1 is regressed with the spectrally similar Landsat band **4, SPOT** band **2** with Landsat band **5,** and **SPOT** band 3 with Landsat band 7. The regression equation for each band pair is of the form

$$
\hat{\mathbf{C}}_{Li} = \frac{\mathbf{C}_{Si} - \mathbf{B}_{i}}{\mathbf{M}_{i}} \tag{1}
$$

where

 \hat{C}_{Li} = approximated Landsat count value for band i;

 C_{Si} = **SPOT** simulator count value; and M_i and $\ddot{B_i}$ = regression coefficients.

The regression coefficients generated for each band pair are shown in Table 1.

Before applying the unsupervised classifier, all land areas are masked by choosing a threshold value within the infrared band; **SPOT** simulator band 3 or Landsat band 7. Pixels with count values greater than the threshold value are assigned a value of **255** within all bands. This procedure causes the classifier to use one class to represent all land pixels, thereby forcing all other classes to be assigned to water features.

The unsupervised classifier consists of a clustering routine developed by ERDAS, Inc. The user is required to enter the maximum number of classes to be created (N) , the minimum distance allowed between cluster centers **(C),** and the maximum radius of each cluster (R). The values selected for N, **C,** and R are the only control the user has over the program. Because it is the classification that will be used to compare data types, the derivation of N, **C,** and R must be consistent. We begin by defining $N = 31$. This value is selected because it is the largest number of classes that can be easily manipulated on the ERDAS **400.**

The minimum distance between clusters and maximum cluster radius are both distance measures

TABLE 1. LIXEAR RECRESSION COEFFICIENTS CALCULATED FOR RADIOMETRICALLY **DEGRADING THE SPOT** SIXILLATOR DATA

Band Pair (SPOT:Landsat)	M_{\cdot}	В (Counts)	
1:4	3.75	54.84	0.96
2:5	2.49	52.01	0.98
3:7	2.18	16.48	0.98

in spectral space. We define these values as $C = D/31$ and $R = C/2$ where

$$
D = [(L_b - L_d)^{T} (L_b - L_d)]^{1/2}
$$
 (2)

and L_b and L_d are mean count value vectors for bright water and dark water, respectively.

Upon completion of the classification program, the **31** classes are grouped into five categories:

- **Land**
- **Marsh Water Plume**
- **Tidal Creeks**
- **Nearshore Bay Water**
- Offshore **Bay Water**

This is accomplished by overlaying each class onto the raw image and comparing the distribution with aerial photography collected during the **SPOT** simulator overflight. The aerial photography is **1:42,000** scale color infrared transparencies showing the entire area imaged in the **SPOT** simulator data.

RESULTS

The **SPOT** simulator data were collected during a time of day when the sun was high in the western sky; altitude **54",** azimuth **254".** Consequently, water features in the western half of the image are totally obscured by sun glint. Data analysis is therefore confined to the central and eastern half of the image where surface reflection is minimized.

Both the **SPOT** simulator and Landsat-3 MSS data indicate a plume of spectrally unique water emanating from Roosevelt Inlet. The plume is most evident in **SPOT** band 1 **(500-590** nm) and the spectrally similar Landsat **MSS** band **4 (500-600** nm). In each case, the plume appears dark relative to the surrounding bay water. Both images also indicate a decrease in volume reflectance from the ambient bay water with increasing distance from shore. In **SPOT** band **2 (610-680** nm), the plume appears less visible, although it still appears darker than the ambient bay water with distance from shore.

Results of the unsupervised classification are shown in Figure 3. In Figure 3a, the raw **SPOT** data clearly show a plume emanating from Hoosevelt Inlet and progressing northwest along the bay shoreline. Spectrally similar water is shown located offshore, but this is believed to be an effect of sun glint. The classification shows good separation of water masses with little confusion between classes. Contrary to these results, the classified Landsat **MSS** data, shown in Figure 3b, indicates confusion between classes. Although a plume is suggested emanating from Roosevelt Inlet and progressing southeast along the bay shoreline, numerous pixels located offshore are also classified as plume, suggesting confusion with the two bay classes. Neither the plume, nearshore bay, or offshore bay classes are as spatially homogeneous as in the classified raw **SPOT** data. Applying the classifier to the radiometrically degraded **SPOT** data resulted in class

FIG. 3. Cluster analysis, Canary Creek.

distributions and homogeneity surprisingly similar to those of the raw **SPOT** data. This implies that the decreased radiometric sensitivity of the Landsat **MSS** sensor data does not, in itself, account for the apparent confusion among water masses.

The spectral distance parameter (D) is calculated between classes for each Canary Creek classification. The results are presented in Table 2. Values of D are proportional to degrees of separability of classes. The raw **SPOT** simulator data resulted in values significantly higher than the radiometrically degraded **SPOT** data and the Landsat **MSS** data. In all three cases, plume water is most similar to nearshore bay water. This is where most of the confusion occurs within the Landsat classification. The radiometrically degraded **SPOT** data resulted in D values similar to those representing the Landsat classification. This appears inconsistent when comparing the amount of confusion between water masses within the two classifications. However, upon comparing the two unclassified images, it is evident that a significant amount of banding is present within the Landsat data. Because the **SPOT** data have no

banding and lower levels of sensor noise, a D value between classes can be quite small, yet the classes remain separable. It is anticipated that banding will not be a problem within the operational **SPOT** data because of the linear array configuration of the sensor.

The results of the unsupervised classifier applied to the Roosevelt Inlet images, shown in Figure 4, are much the same as in the previous discussion. The **SPOT** data, both raw and radiometrically degraded, produced water classes that were well-defined and spatially homogeneous. However, because the **SPOT** data are now at 20-metre resolution, the classifications show more detail. The 60-metre **SPOT** data indicate a change in volume reflectance along the axis of the Broadkill River, while the 20 metre data show variations across the river. The outline of tidal creeks are more clearly shown in the higher resolution data, although the water classes here show various degrees of edge effect. Once again, the Landsat data suggest confusion between classes.

The spectral distance parameter is calculated be-

FIG. 4. Cluster analysis, Roosevelt Inlet.

tween classes for each Roosevelt Inlet classification and the results are shown in Table 3. Again, the raw REFERENCES **SPOT** data produced the largest values for all class **Ackleson, S.** C., 1981. *The Verification of a Radiatiue* pairs, indicating superiority in separating the spec-

SPOT SIMULATOR **tral signatures of water masses**. The smallest value of D was calculated between the plume and offshore bay classes. Values of D between these classes representing radiometrically degraded **SPOT** data and Landsat **MSS** data were quite similar, 2.24 and 2.32, respectively. However, the manner in which apparent confusion occurs is different. Within the Landsat image, pixels classified as plume occur throughout the offshore bay area, while pixels classified as offshore bay occur throughout the plume region. Most often the misclassified pixels are oriented along diagonal lines similar to the banding. Within the radiometrically degraded **SPOT** data, there is evidence of confusion only along the boundary of these two classes. Pixels in this area are spectrally similar and the higher radiometric sensitivity of the raw **SPOT** data is required to separate the two classes. In the Landsat data, this effect is undoubtedly occurring, but sensor noise appears to be the dominant cause of misclassification.

CONCLUSIONS

SPOT simulator and Landsat **MSS** data are shown to be effective in delineating water masses in Delaware Bay and, to a lesser extent, Broadkill River. The **SPOT** simulator holds several advantages over Landsat **MSS** data:

- **The greater radiometric sensitivity of** SPOT **data produces better separation of spectrally similar water masses,**
- **The** SPOT **data contain less noise resulting in fewer ~nisclassified pixels, and**
- **The higher ground resolution of** SPOT **data provides indications of small-scale hydrodynamics not pos**sible within Landsat data.

SPOT imagery promises to be of significant value to the estuarine and coastal hydrographer. It is anticipated that the most useful attribute of the data will be the 20-metre resolution of the multispectral data and the 10-metre resolution of the panchromatic band. Here, the data will aid the researcher in understanding dynamic processes in surface waters such as estuarine circulation and detrital flux

Application to Remotely Sensed Imagery: Master's Thesis, University of Delaware, Newark, Delaware, 141 p.

- DeWitt, Piet, and Frank C. Daiber, 1973. The Hydrography of the Broadkill River Estuary, Delaware: *Chesapeake Science,* Vol. *14,* No. 1, pp. 28-40.
- Klemas, V., M. Otley, W. Philpot, C. Wethe, R. Rodgers, and N. Shah. 1974. Correlation of Coastal Water Turand N. Shah, 1974. Correlation of Coastal Water Tur-

bidity and Current Circulation with ERTS-1 and aware Salt Marsh: Doctoral Dissertation, University Skylab Imagery: *Proceedings of the Ninth Interna-*

tional Symposium on Remote Sensing of Environment. Ann Arbor, Michigan, pp. 1289-1317.

- Klemas, V., E C. Daiber, W. D. Philpot, S. G. Ackleson, and C. T. Roman, 1981. *Remote Sensing of Organic Inorgarlic Material in Coastal and Estuarine Waters:* Final Sea Grant Report R/B-7, College of Marine Studies. Universitv of Delaware, Newark, Delaware,
- aware Salt Marsh: Doctoral Dissertation, University
of Delaware, Newark, Delaware, 144 p.

ANNOUNCEMENT AND CALL FOR PAPERS

Remote Sensing Applied to Monitoring and Management of Wetlands

Louisiana State University, Baton Rouge 4-5 October 1985

This Annual Fall Meeting of the Mid-South Region is being organized by the Mid-South Region of the American Society for Photogrammetry and Remote Sensing and the Remote Sensing and Image Processing Center of the Louisiana State University. It will be preceded by a one-day workshop on Photointerpretation of Wetlands on 3 October, conducted by LSU faculty and specialists from the USGS.

Those wishing to present papers on "Remote Sensing Applied to Monitoring and Management of Wetlands" should subinit a 200-word abstract before 1 *Atrgust* 1985 to

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INVITATION AND CALL FOR PAPERS

Tenth Canadian Symposium on Remote Sensing

Edmonton, Alberta, Canada 5-8 May 1986

This Symposium-sponsored by the Canadian Remote Sensing Society of the Canadian Aeronautics and Space Institute, the Canadian Institute of Surveying, and the Canada Centre for Remote Sensingwill feature all aspects of Remote Sensing, including

- Sensors
- Data Acquisition
- **•** Processing and Analysis
- Environmental Monitoring

with special emphasis on

Value of Remotely Sensed Data in Operational Use.

Those interested in presenting a paper should submit a 200-word (maximum) abstract by 2 *December* 1985 to

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