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2 REMOTE SENSING PROVIDES A NEW LOOK AT THE  
3 EARTH AND ITS RESOURCES<sup>1</sup>

4  
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6  
7 INTRODUCTION

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9 In 1948 the British astrophysicist Fred Hoyle wrote:

10 "Once a photograph of the Earth, taken from the outside, is  
11 available--once the sheer isolation of the Earth becomes plain-- a  
12 new idea as powerful as any in history will be let loose".

13 In 1977 Robert Frosch, administrator of the U. S. National  
14 Aeronautics and Space Administration (NASA), observed:

15 "This is the first generation of people that has actually seen  
16 the Earth. What men saw before were only little bits and pieces of  
17 the Earth. Man would take these little bits and pieces and hang  
18 them together in maps in an attempt to construct a picture of the  
19 Earth as it would be seen from space. A few decades ago he invented

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1 the camera and later placed it on airplanes. He could then get bigger  
2 pictures, and less piecing together was necessary. But this is the  
3 first generation that has seen data flow in the opposite direction--  
4 data that began with a broad synoptic view from which one then has to  
5 extract the details. In a sense, we have turned the whole enterprise  
6 around. Instead of starting with the details and trying to construct  
7 the big picture, we now have the capability to go the other way--to  
8 look at the big picture and then learn how to extract the details  
9 which explain it" (Frosch, 1977).

10 Modern engineering technology has given our generation the first  
11 opportunity to view the Earth as a whole. Dramatic advances in  
12 remote sensing technology make it possible not only to survey Earth  
13 surface features with a high degree of detail and accuracy but also  
14 to monitor changes occurring on the Earth surface.

#### 15 16 NEW TECHNOLOGY FOR OBSERVING THE EARTH

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18 Advances in three areas of technology during the past twenty-five  
19 years have made possible the development of repetitive Earth observa-  
20 tion systems. These areas of new technology involve a sensor system,  
21 computers, and communications.

22 1. Sensor systems. During recent years a broad array of instruments  
23 have been developed for observing and measuring specific components  
24 or features of our environment. These possibilities range from a

1 look at the interior of the atom to a synoptic, repetitive color view  
2 of the entire surface of the Earth; from a sub-microscopic examination  
3 of plant cells under moisture or disease stress to a view of thousands  
4 of hectares of wheat suffering from drought. The wide array of new  
5 instruments and analytical techniques allows us to observe the same  
6 environment at different scales and levels of detail.

7 2. Computers. Another new capability is provided by the electronic  
8 computer without which many of the new data acquisition instruments  
9 could not function. The computer provides data storage, retrieval  
10 and analytical capabilities unthinkable even a decade ago. The age  
11 of the electronic, digital, stored program computer began in the early  
12 1950s (Davis, 1977). A profound admiration of and interest in com-  
13 puters by people is to be expected. Computers are properly cited as  
14 the first as well as the most important invention ever that signif-  
15 icantly extends human intellectual capabilities. Davis (1977)  
16 suggests that until the age of computers, inventions had primarily  
17 extended human muscular powers as well as certain sensory powers.  
18 Now within the past 25 to 30 years there has descended upon us an  
19 electronic array of big computers, super computers, "intelligent"  
20 terminals, industrial robots, minicomputers, microcomputers, computers-  
21 on-a-chip and tele-operators. This technology provides the opportunity  
22 for a conceptual framework for resource information systems inconcei-  
23 vable a generation ago.

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1 3. Communications. A third development of significance to resource  
2 information systems is in the area of communications. Present capa-  
3 bilities of communications satellites permit the high quality, instan-  
4 taneous transmission of verbal messages, images and mass quantities  
5 of data from any location to any other location on the surface of  
6 the Earth. A remarkable outgrowth of this new communications tech-  
7 nology has been the organization and operation since 1964 of  
8 INTELSAT, an international organization consisting of membership of  
9 more than one hundred signatory nations. INTELSAT's prime objective  
10 is the provision, on a commercial basis, of the space segment required  
11 for international public telecommunications service of high quality  
12 and reliability, to be available on a non-discriminatory basis to all  
13 areas of the world (INTELSAT, 1978).

#### 14 15 AN INFORMATION SYSTEM FOR EARTH RESOURCES

16  
17 Information is a valuable commodity. One of the often overlooked  
18 features of the development and management of resources--agricultural,  
19 forestry, mineral--is the supporting information system. As the  
20 demands increase for greater and more efficient production of food  
21 or more efficient utilization of dwindling resources, the role of  
22 information becomes more critical. It becomes more important that  
23 accurate, useful, inexpensive and timely information be available to  
24 the decision-makers and policy-makers. Accurate and timely

1 information must be available to policy-makers so that rational  
2 decisions may be made in the allocation of limited resources for  
3 development. The efficiency of a nation's agriculture may be related  
4 to the quality and quantity of information available to decision-  
5 makers and policy-makers.

6 The lack of accurate and timely information about natural  
7 resources--soils, land capability, land degradation, areas and con-  
8 ditions of crops, quantity and quality of water resources, quantity  
9 and conditions of forests and rangelands, rates of change in land  
10 use--places great limitations on resource managers and policy-makers  
11 in making rational decisions.

12 It is now technically possible to design and implement an Earth  
13 resources information system supported by Earth orbiting satellite  
14 sensors. Data from such a system could be used to derive a current  
15 inventory of land, crop, forest, rangeland and surface water resources.  
16 Further, regular sequential scanning of these resources could provide  
17 a quantitative assessment of seasonal and yearly changes which occur  
18 in the landscape.

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## CURRENT EARTH OBSERVATION SATELLITE SYSTEMS

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4       The decade of the 1970's introduced the world to the concept of  
5 observing the Earth from space. During this decade much has been  
6 learned about collecting data from space, extracting useful informa-  
7 tion from space-derived data, and interpreting and using such infor-  
8 mation.

9   1. Data acquisition. Modern remote sensing techniques are based upon  
10 the fact that information is conveyed from the Earth to the sensor in  
11 force fields and electromagnetic fields emanating from the scene and  
12 and in particular through the spectral, spatial and temporal varia-  
13 tions of such fields (Figure 1). In order to derive information  
14 about the scene, one must be able to measure these field variations  
15 and relate them to scene characteristics or Earth surface features of  
16 interest.

17       Although remote sensing techniques are being used effectively  
18 near ground level in the field and at low and high altitudes with  
19 aerial sensors, the primary objective of this paper is to describe  
20 the characteristics and use of data derived from current Earth  
21 observation satellites.

22       In July 1972 the National Aeronautics and Space Administration  
23 (NASA) launched the first Earth observation satellite with instruments  
24 designed to observe the land areas of the globe. Known as Landsat 1,

1 this satellite was placed in a near polar sun-synchronous orbit at an  
2 altitude of approximately 920 km. Landsat 1 was equipped with an  
3 array of three return beam vidicon (RBV) cameras and a four-band  
4 multispectral scanner (MSS). The orbital design was such that  
5 Landsat 1 had the capability of scanning the entire surface of the  
6 Earth every eighteen days.

7 Landsat 2 with specifications identical to those of Landsat 1  
8 was launched in January 1975. Landsat 3 was placed into orbit in  
9 March 1978. In addition to the four MSS bands of Landsats 1 and 2,  
10 the scanner of Landsat 3 had a thermal infrared band. Improved  
11 spatial resolution was built into the one-band RBV of Landsat 3.  
12 Specifications of the sensors of Landsats 1,2, and 3 are summarized  
13 in Table 1.

14 After five years of successful transmission of Earth resources  
15 data, Landsat 1 was retired from service in January 1978. Deterior-  
16 ation in performance of Landsat 2 in late 1979 led to a shutdown of  
17 its data acquisition system after almost four years of service.

18 Landsat D, which will become Landsat 4 after a successful launch,  
19 is scheduled to be placed in polar orbit during the fourth quarter  
20 of 1981. The two on-board sensors (MSS and thematic mapper, TM)  
21 will provide ground coverage of approximately 31,450 square kilometers  
22 per scene. The MSS will obtain data in the same four reflective bands  
23 as those of previous Landsats and have the same instantaneous field  
24 of view (80 m). The TM, a line-scanning device also, will operate

1 over seven bands of the visible and infrared range of the electro-  
2 magnetic spectrum and have an instantaneous field of view (IFOV)  
3 of 30 meters for all reflective bands (Table 2).

4 A comparison between the orbital and scanning characteristics of  
5 Landsat D and the first three Landsats is presented in Table 3.

6 When Landsats 1, 2 and 3 are beyond the reception range of a  
7 receiving station all sensor data must be recorded on-board the  
8 satellite and transmitted later when the space vehicle moves within  
9 the range of a station. Data volume constraints of the recording  
10 tape and technical problems with the on-board recording systems have  
11 seriously limited the capability of the Earth observation satellites  
12 to obtain data regularly over land areas outside the reception range  
13 of ground stations. The common reception radius of ground receiving  
14 stations is 2,780 km.

15 Data from Landsat sensors are transmitted to receiving stations  
16 in digital form. Receiving stations in operation as of October 1979  
17 include the following:

18 United States of America (3 stations)

- 19 - Greenbelt, Maryland  
20 - Goldstone, California  
21 - Fairbanks, Alaska  
22  
23  
24



1 Canada (2 Stations)

2 - Prince Albert, Saskatchewan

3 - Shoe Cove, Newfoundland

4 Brazil - Cuiaba

5 Italy - Fucino

6 Japan - Tokyo

7 Sweden - Kiruna

8 Receiving stations are under construction in Argentina,  
9 Australia and India.

10 The data products available from receiving stations include  
11 single-band black and white images, color composite (multi-band)  
12 images, and digital data on computer compatible tape (CCT). These  
13 images can be analyzed and interpreted by visual methods, or the  
14 digital data can be analyzed by computer-implemented pattern recog-  
15 nition techniques.

16 2. Data analysis. Traditionally the ability of the observer to  
17 acquire data has exceeded by far his or her ability to analyze and  
18 interpret data. The computer revolution of the past twenty-five  
19 years has provided a major advance in the human capability to cope  
20 with the masses of data which are accumulated. Computers, or hard-  
21 ware, together with computer languages and analytical programs, or  
22 software, which have evolved during the past two decades, place at  
23 our disposal for the first time a capability to store, retrieve,  
24 overlay, analyze and-interpret vast quantities of data.

1 Computer technology is essential for handling and analyzing the  
2 masses of data acquired by the sensors of Landsats 1, 2, and 3, each  
3 of which was designed to transmit more than one million quantitative  
4 reflectance measurements per second. In the analysis of Landsat  
5 data computers may be used to produce high quality images for visual  
6 analysis by the human interpreter (Figure 6) or to analyze the data  
7 by pattern recognition techniques for producing a classification of  
8 landscape features of a specific area (Figure 7). Results of such  
9 analyses may be presented as a map, as spectral curves, and/or in  
10 tabular format indicating quantitatively the area classified as wheat,  
11 maize, poorly drianed soil, sand, coniferous forest or other landscape  
12 features of interest.

13 In this paper emphasis will be given to the application of pat-  
14 tern recognition techniques by digital analysis of multispectral data.  
15 The principle of computer-implemented pattern recognition techniques  
16 may be illustrated by examination of the spectral properties of  
17 different features in the landscape. Typical spectral curves have  
18 been plotted for green vegetation, bare soil and water (Figure 2).  
19 The horizontal axis represents wavelength in the visible (0.38 to  
20 0.72  $\mu\text{m}$ ) and reflective infrared (0.72 to 2.50  $\mu\text{m}$ ) portions of the  
21 electromagnetic spectrum. The vertical axis represents the intensity  
22 of reflected energy as measured by a spectroradiometer. An examin-  
23 ation of these curves reveals that there are certain wavelengths that  
24 are much better than others in separating green vegetation, soil and

1 water.

2 The same data used for plotting these spectral curves may be  
3 plotted in another fashion. For example, the reflectance values for  
4 each landscape feature (vegetation, soil, water) at three different  
5 wavelengths, represented by  $\lambda_1$ ,  $\lambda_2$ ,  $\lambda_3$ , may be plotted in three  
6 dimensional space (Figure 3).

7 In the analysis of multispectral data by pattern recognition  
8 techniques, one of two general approaches is usually followed.

9 One approach is termed "clustering" or "non-supervised" classifi-  
10 cation. With this approach an algorithm is used which directs the  
11 computer to examine the spectral data for the area of interest and to  
12 assign each picture element (pixel) to a cluster of pixels having sim-  
13 ilar spectral characteristics. The number of cluster classes to be  
14 spectrally separated is generally set arbitrarily by the analyst and  
15 may or may not be determined by the analyst's prior knowledge of the  
16 area being analyzed.

17 The other general approach is termed "supervised" classification.  
18 In this case the analyst provides the computer with a spectral defini-  
19 tion of the classes to be spectrally separated. This spectral defini-  
20 tion is provided in the form of a set of training samples of known  
21 identity from specific addresses within the multispectral data. The  
22 quality of the supervised classification results is dependent upon the  
23 spectral separability of the desired classes with existing spectral  
24 data and upon the quality or representativeness of the training set

1 selected by the analyst.

2       The identification and delineation of green vegetation, bare soil  
3 and water by spectral analysis is quite simple and can be accomplished  
4 with a high level of accuracy with little or no knowledge of the  
5 ground scene by the analyst. However, the classification problem by  
6 spectral analysis becomes much more difficult when the objective is to  
7 classify the ground scene into many different subcategories of green  
8 vegetation, soil and water. Subcategories of interest might include  
9 many crop species, differences in plant vigor and biomass production,  
10 soils with varying contents of organic matter or different internal  
11 drainage characteristics, and water containing varying amounts of sus-  
12 pended matter. Complexity of classification at this level requires  
13 special interpretive skills by the analyst. Such skills can best be  
14 obtained through broad experience in spectral analysis and intimate  
15 knowledge of the area being classified.

16       Although human experience is limited to three-dimensional percep-  
17 tion, the computer has no such constraints; it can operate in n-dim-  
18 ensional space. Thus, with appropriate algorithms the computer can  
19 examine the quantitative reflectance values for each of the four wave-  
20 length bands of the Landsat scanners and classify or assign each data  
21 point or picture element(pixel) to a specific spectral category.  
22 Many different algorithms have been developed for the computer-imple-  
23 mented analysis of multispectral data.

24       The remainder of this paper will be a presentation of

1 applications of the digital analysis of remotely sensed data.

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### USES OF LANDSAT DATA

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5 Since the launch of Landsat 1 in July 1972, scientists in many  
6 nations have used satellite data to inventory and monitor land,  
7 mineral, vegetation and water resources. Case studies discussed in  
8 this paper will be limited to applications related to agricultural  
9 production.

10 1. Soil survey. During the 1970s, scientists in increasing numbers  
11 have been exploring methods of using satellite-derived images for  
12 delineating meaningful soil boundaries and characterizing soil con-  
13 ditions. One of the important uses of Landsat images is as a base  
14 from which preliminary soil legends can be described and general soil  
15 differences can be mapped. A single satellite image has a particular  
16 advantage over aerial photography because it provides a synoptic view  
17 of a contiguous area covering 34,000 km<sup>2</sup>. This synoptic view is  
18 valuable to the soil surveyor in correlating soils mapped in adjacent  
19 counties or political units by different surveyors at different times.

20 Visual interpretation of black and white and color composite  
21 images produced from Landsat data were used by Westin and Frazee  
22 (1976) to produce general soil maps at scales of 1:250,000 or smaller  
23 with map units of 260 hectares or larger. Weismiller et al (1979)  
24 visually interpreted black and white and color composite Landsat

1 images to determine soil parent material boundaries in Jasper County,  
2 Indiana (Figure 4 ). Confirming these boundaries by field observa-  
3 tions, they then digitized these boundaries and overlaid them onto  
4 the digital data of the four Landsat MSS reflectance bands which had  
5 been adjusted to a scale of 1:15,840. Spectral maps delineating soil  
6 differences were then produced at this scale for the entire county  
7 (1453 km<sup>2</sup>) with map units as small as one hectare. These maps delin-  
8 eating more than fifty different spectral classes in six different  
9 parent material areas are being used by soil surveyors to supplement  
10 conventional black and white aerial photographs in the preparation  
11 of a detailed soil survey of Jasper County (Figure 5 ). Soil survey-  
12 ors have observed that satellite-derived spectral maps can make a  
13 significant contribution in improving the accuracy of the final soil  
14 map and in decreasing the time required to produce a detailed soil  
15 survey.

16 The digital format of Landsat MSS data permits rapid and easy  
17 merging and/or recombination of spectral classes to produce a wide  
18 array of smaller scale base maps for soil surveys at different levels  
19 of generalization.

20 2. Land degradation. There is increasing evidence that Landsat data  
21 can be useful for mapping and monitoring land degradation caused by  
22 wind erosion, water erosion, salinization and flooding. A striking  
23 example of denudation and sand dune encroachment caused by wind  
24 action can be seen in an area along the Wadi Abu Habl in Western

1 Sudan (Figure 6 ). The same data from which this synoptic view of  
2 the area at a scale of 1:250,000 was produced, was also used to exa-  
3 mine in detail at a scale of 1:25,000 the characteristics of the cul-  
4 tivated areas of one of the sand dunes (Figure 7 ). The three  
5 spectral classes of the cultivated sands seemed to correlate well  
6 with the areas of millet, peanuts and fallow.

7 Severe erosion caused by rainfall has been detected and separated  
8 spectrally in the humid temperate regions of northern and central  
9 Indiana (Figure 8 ). In these cultivated soils the exposure of  
10 subsoil resulting from severe erosion gives reflectance values measur-  
11 ed by the satellite sensors which are considerably different from  
12 those of the less eroded surrounding soils (Figure 9 ).

13 3. Crop inventory. One of the most obvious agricultural applications  
14 of remote sensing technology is the identification and area measure-  
15 ment of cultivated crops (NASA, 1978). Equally important, but some-  
16 what more complex is the use of this technology to improve crop  
17 yield predictions. An increasing research effort is being expanded  
18 to use remotely sensed data to provide essential inputs into crop  
19 yield prediction equations.

20 For many years government agencies in some countries have used  
21 aerial photography to identify crop species and measure the areas  
22 planted to specific crops. This is a rather expensive and time-con-  
23 suming method to conduct crop inventories which may change from year  
24 to year.

1       The need to have more accurate and timely information about  
2 global grain supplies and food crop conditions has intensified during  
3 the past two decades. The Earth Observation Satellite program has  
4 opened new possibilities for inventorying the major food crops of the  
5 world and for monitoring crop conditions periodically through the  
6 growing season.

7       An example of the use of digital analysis of Landsat data to  
8 identify and measure the area of wheat is illustrated in Figure 10.  
9 In this example spectral analysis was used to identify three features  
10 in the landscape of Greeley County, Kansas. These features were  
11 fields of winter wheat, fields prepared for seeding summer crops and  
12 rangelands generally associated with the rough terrain along natural  
13 drainageways. In this county with a total land area of 202,803  
14 hectares, the U.S. Department of Agriculture area estimate for winter  
15 wheat in 1973 was 73,000 hectares  $\pm$  5%. This estimate was derived  
16 from a standardized field sampling scheme which examines no more than  
17 two percent of the land area of the county. The 1973 wheat estimate  
18 for the county derived from data obtained by a Landsat 1 pass on  
19 19 June 1973 was 77,000 hectares. In this instance with Landsat data  
20 every hectare of the county was examined and analyzed spectrally.

21       Early results such as these led to the planning and implementa-  
22 tion of the Large Area Crop Inventory Experiment (LACIE), a coopera-  
23 tive research program involving the U.S. Department of Agriculture  
24 (USDA), the National Aeronautics and Space Administration (NASA),



1 and the National Oceanic and Atmospheric Administration (NOAA).

2 LACIE research was conducted during the period 1974 to 1978. Its  
3 primary objective was to improve the area, yield and production  
4 estimates of wheat in the winter and spring wheat area of the Central  
5 Great Plains of the United States. Results confirmed that remote  
6 sensing can make a significant contribution in identifying crops and  
7 making area estimates over large regions. Evidence was observed that  
8 remote sensing can be used effectively to monitor gross changes in  
9 crop conditions over large areas.

10 Although the LACIE program was terminated in 1978, new research  
11 programs are being continued and planned for the 1980s. These  
12 programs will expand crop inventory research to include other major  
13 food crops. Much emphasis will be given to the improvement of crop  
14 yield predictions.

#### 15 16 THE FUTURE OF EARTH OBSERVATION PROGRAMS

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18 Throughout the 1970s the Landsat program was officially exper-  
19 imental. In November 1979 the President of the United States issued  
20 a directive assigning the responsibility for civil operational land  
21 remote sensing activities to the National Oceanic and Atmospheric  
22 Administration (NOAA) of the U.S. Department of Commerce. This  
23 provides the long awaited assurance to the international community  
24 that the civil satellite-based Earth resources observation program of

1 the United States will shift from an experimental to an operational  
2 status. This has several important implications for those in the  
3 global community who wish to utilize this technology in the develop-  
4 ment and management of resources. First, it may be interpreted as an  
5 expression of the intent of the United States to continue the Earth  
6 observation satellite program and make satellite data available to  
7 anyone who wishes to purchase it on a non-discriminatory basis.  
8 Second, it suggests that research and development will continue to  
9 support and improve current and future data acquisition and analysis  
10 systems. Third, an operational Earth Observation Satellite system  
11 which can readily make high quality data available to anyone will  
12 greatly accelerate the need for trained personnel to use the tech-  
13 nology. Over the next decade the shortage of persons specialized in  
14 the applications of remote sensing technology may be the single most  
15 serious constraint in its acceptance and use by many nations.

16 If famine is to be averted and global agricultural resources are  
17 to be managed efficiently in the decades ahead, a great improvement  
18 must be made in the acquisition, analysis and delivery of accurate,  
19 essential, useful and timely information to decision-makers and  
20 policy-makers involved in food production. Although remote sensing  
21 is only one of many tools for acquiring and analyzing data, it  
22 promises to become an increasingly important one in the future.

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Table 3 Orbital and Scanning Characteristics  
of Landsats 1, 2, 3 and D

<u>Characteristics</u>	<u>Landsats 1, 2, 3</u>	<u>Landsat D</u>
Altitude	920 km	705 km
Frequency of repetitive coverage	18 days	16 days
Width on ground of scan path	185 km	185 km
Dimension of each Earth scene scanned	185 x 185 km	185 x 170 km

Table 4 Description of lacustrine soils within eight spectral classes derived from Landsat data obtained over Jasper County, Indiana (Figure 5)

<u>Class</u>	<u>Description</u>
1	Excessively to well drained soils with significant inclusions of moderately well drained soils.
2	Excessively to well drained soils with significant inclusions of moderately well drained soils. Minor inclusions of somewhat poorly drained and very poorly drained soils occur.
3	Excessively to well drained soils with a nearly equal amount of moderately well drained soils. Minor inclusions of somewhat poorly drained and very poorly drained soils occur.
4	Very poorly drained soils with significant inclusions of excessively to well drained, moderately drained, and somewhat poorly drained soils.
5	Very poorly drained soils with significant inclusions of somewhat poorly drained soils. Minor inclusions of excessively to well drained and moderately well drained soils occur.
6	Very poorly drained soils with significant inclusions of excessively to well drained and moderately well drained soils.
7	Very poorly drained soils. Minor inclusions of excessively to well drained, moderately well drained and somewhat poorly drained soils occur.
8 (Vegetation)	Predominately moderately well drained soils with significant inclusions of excessively to well drained and very poorly drained soils.

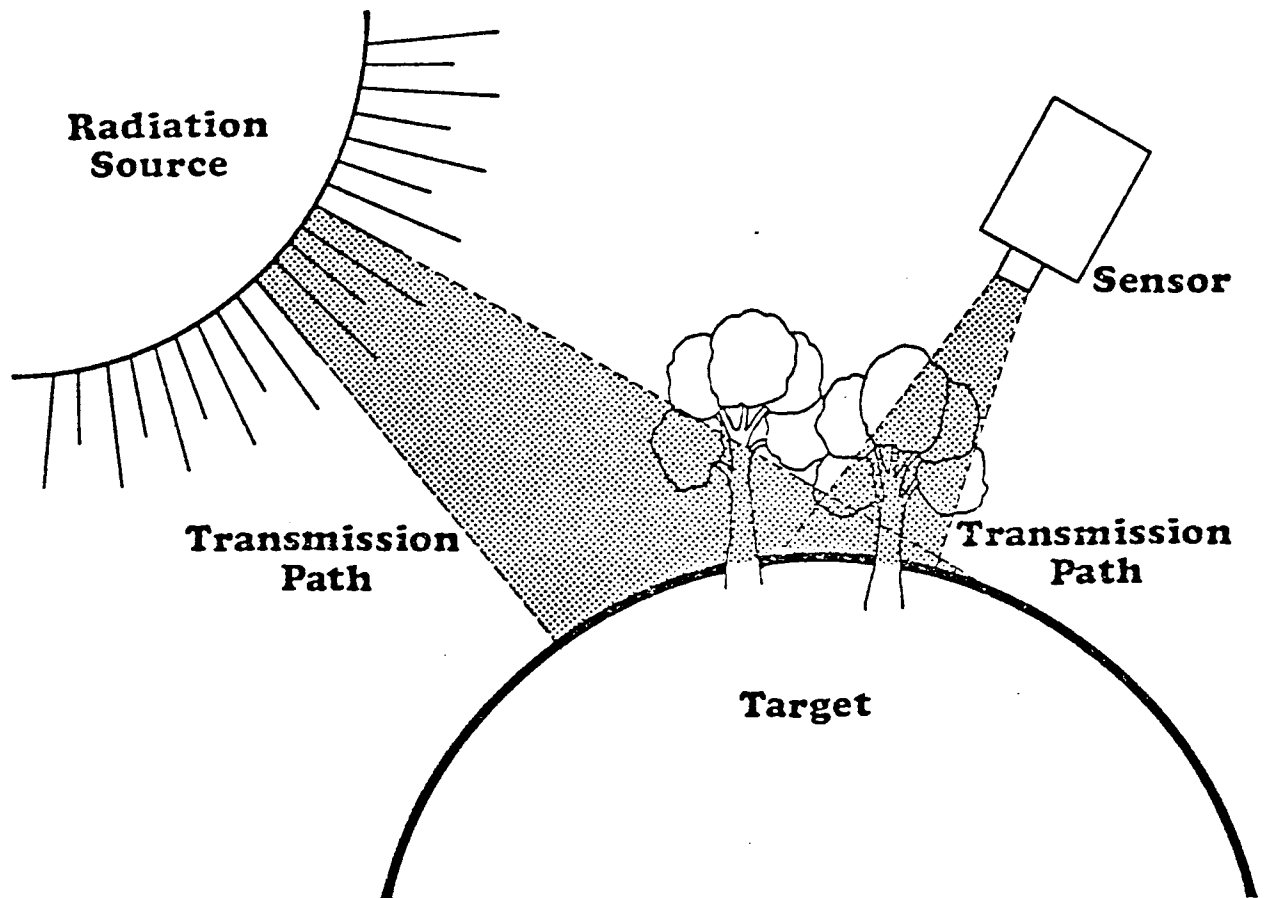


Figure 1. A data acquisition system based on solar radiation and a remote sensor to measure radiation from the scene.

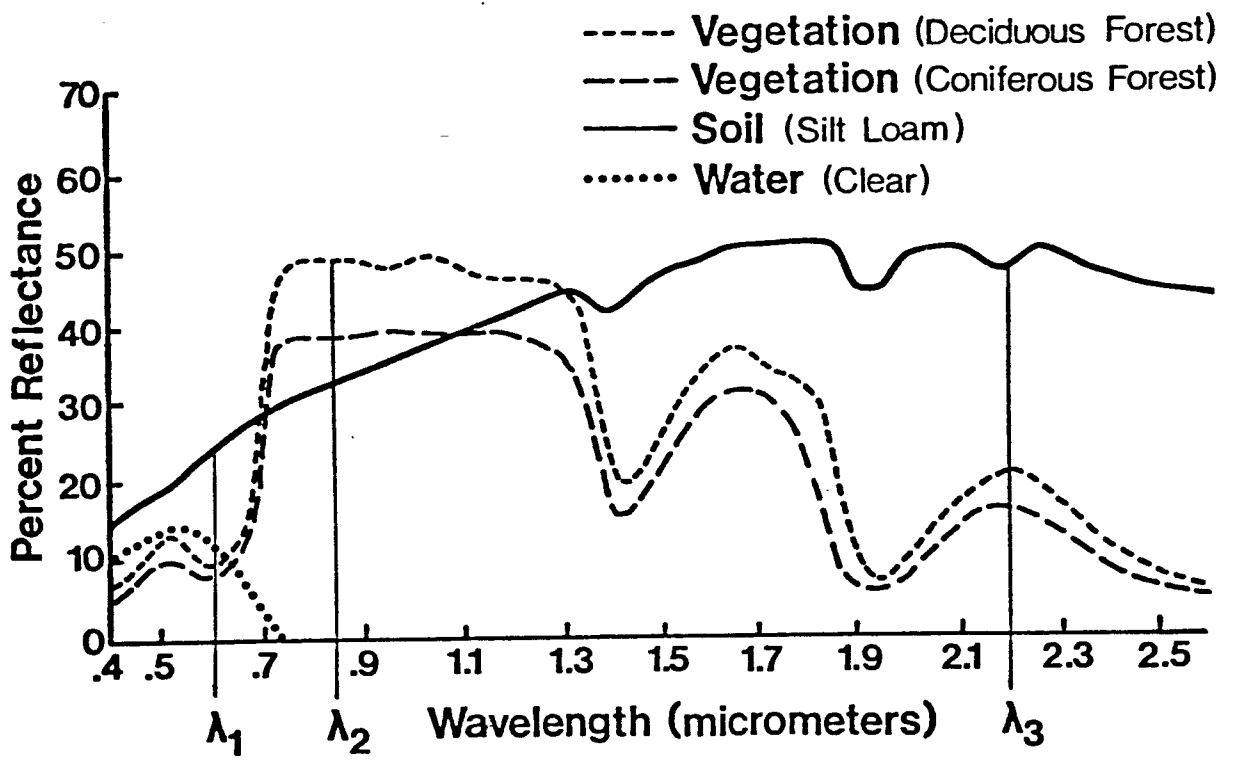


Figure 2. Reflectance curves for green vegetation, soil and water.

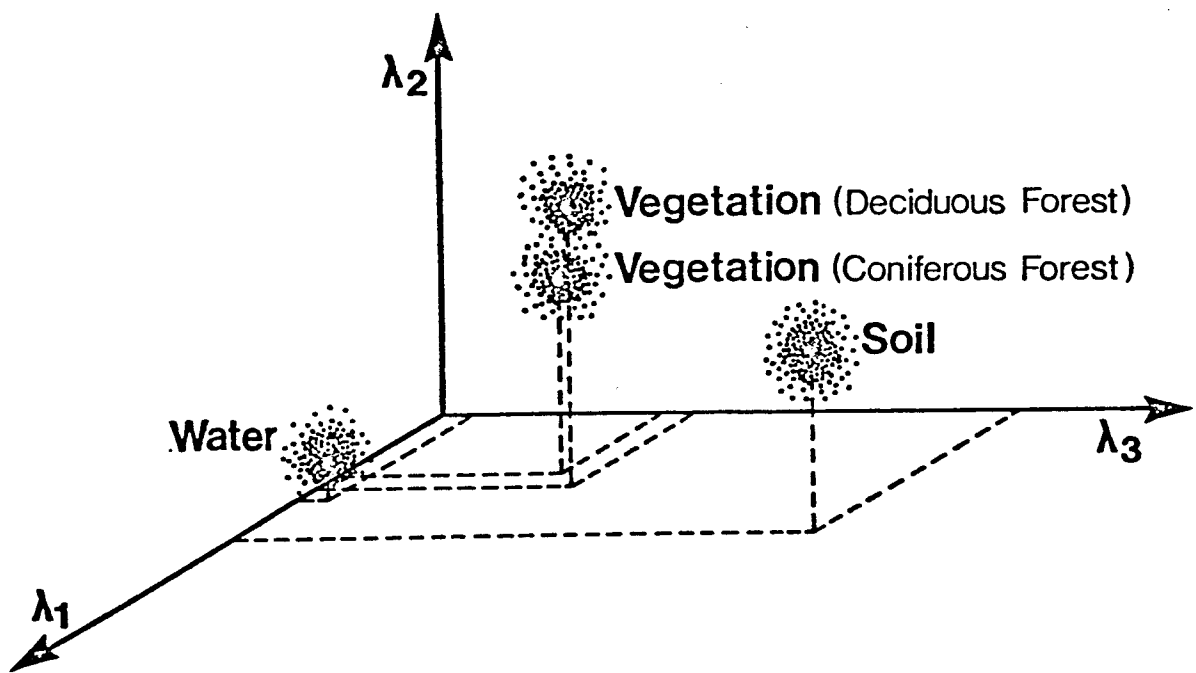


Figure 3. Locating green vegetation, soil and water in three dimensional space defined by specific wavelengths,  $\lambda_1$ ,  $\lambda_2$ , and  $\lambda_3$ .





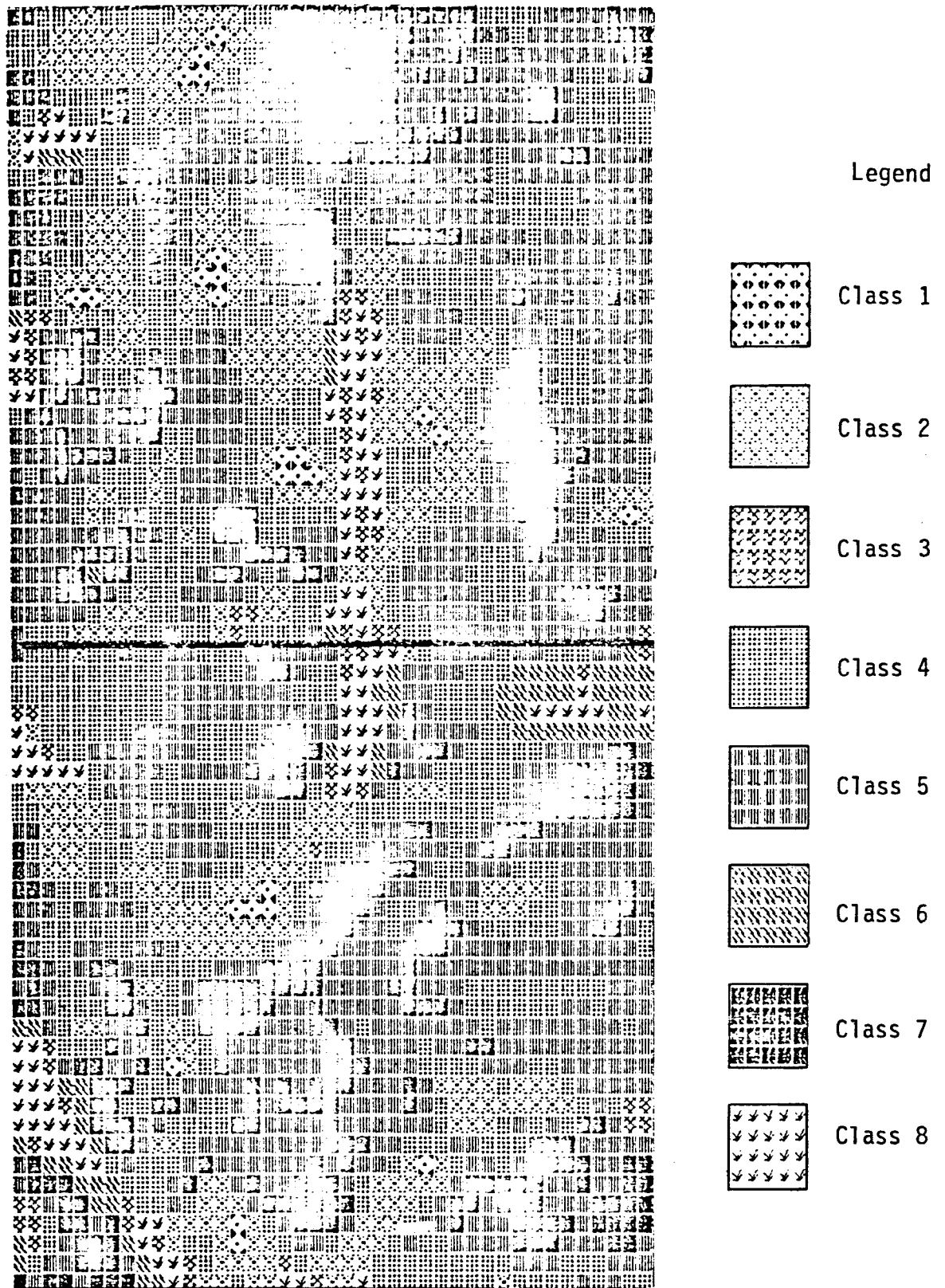


Figure 5. Spectral map derived from Landsat data delineating soil characteristics of a lacustrine area in Jasper County, Indiana. Scale 1:15,840. See Table 4 for class descriptions.



Key to prominent features in landscape:

1. Er Rahad (town)
2. Er Rahad Reservoir
3. Floodplain of Wadi Abu Habl
4. Jebel Ed Dair (granite hill)
5. Upland Clay Plain
6. Umm Ruwaba (town)
7. Sand Dunes
8. Villages

Figure 6. Single-band Landsat image showing denuded areas and dune encroachment along Wadi Abu Habl in Western Sudan. Original scale 1:250,000.

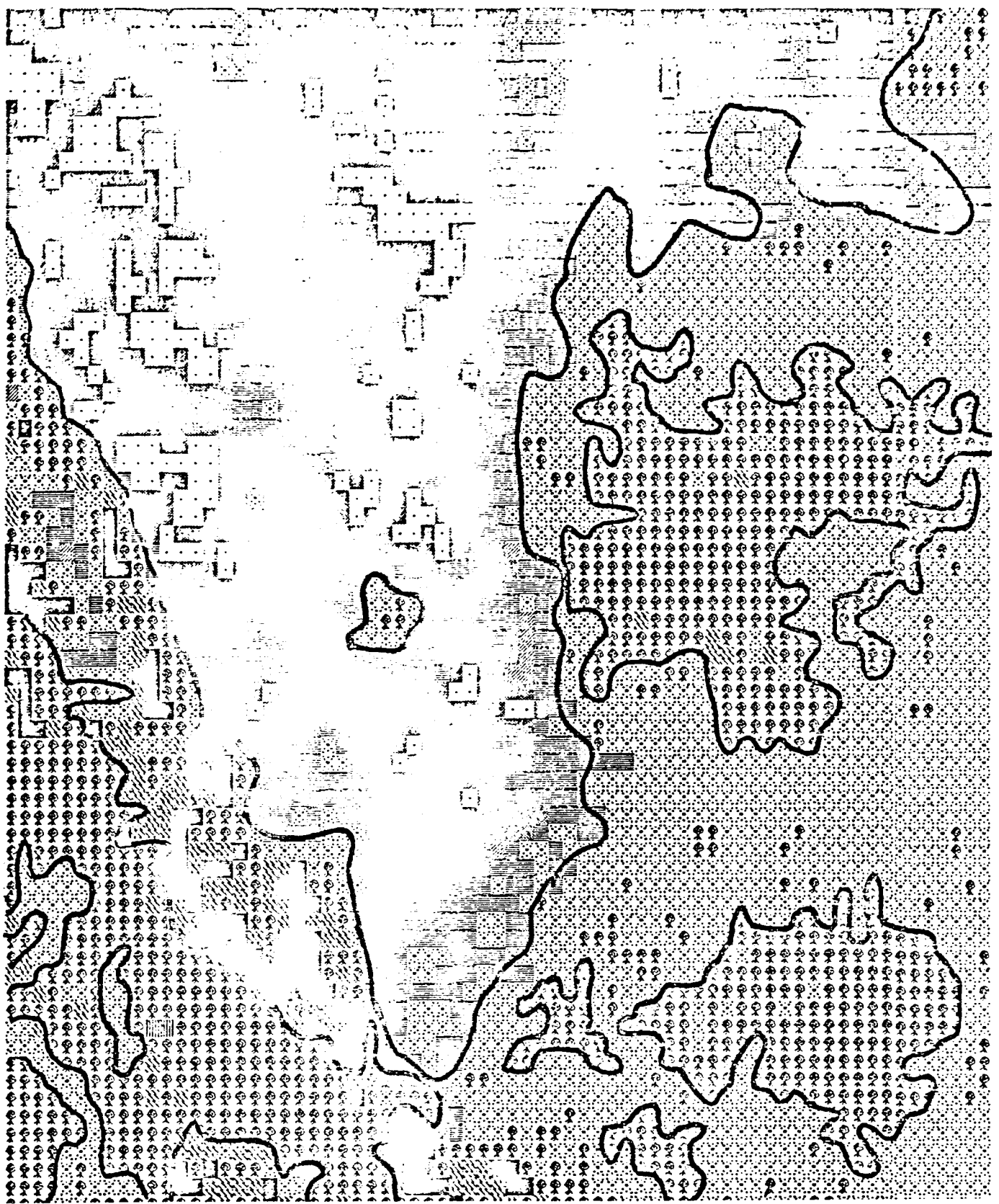


Figure 7. Spectral classification of sand dune area south of Wadi Abu Habl in Western Sudan. Scale: 1:25,000.

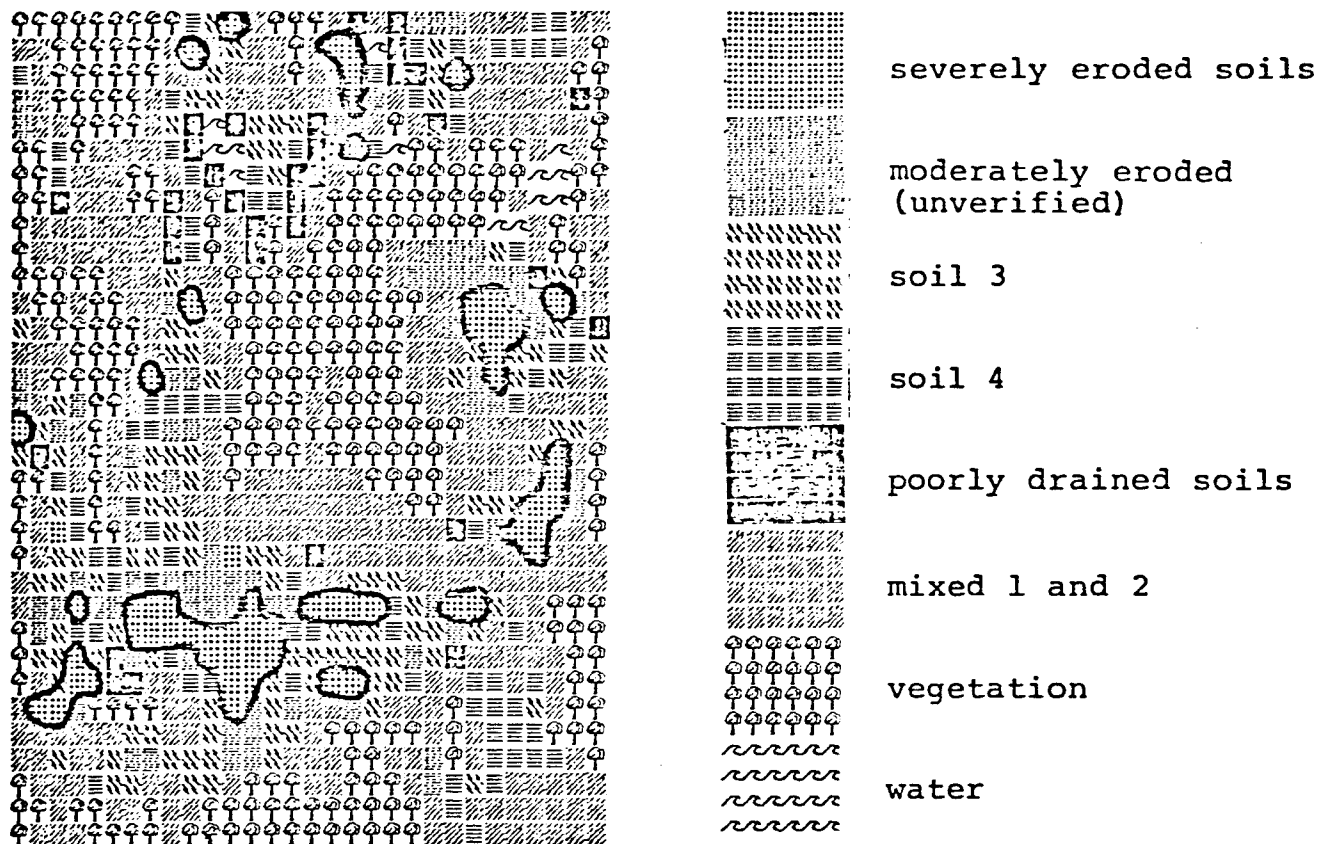


Figure 8. Spectral map units delineating severely eroded soils from other soil and vegetation classes. Scale 1:24,000.

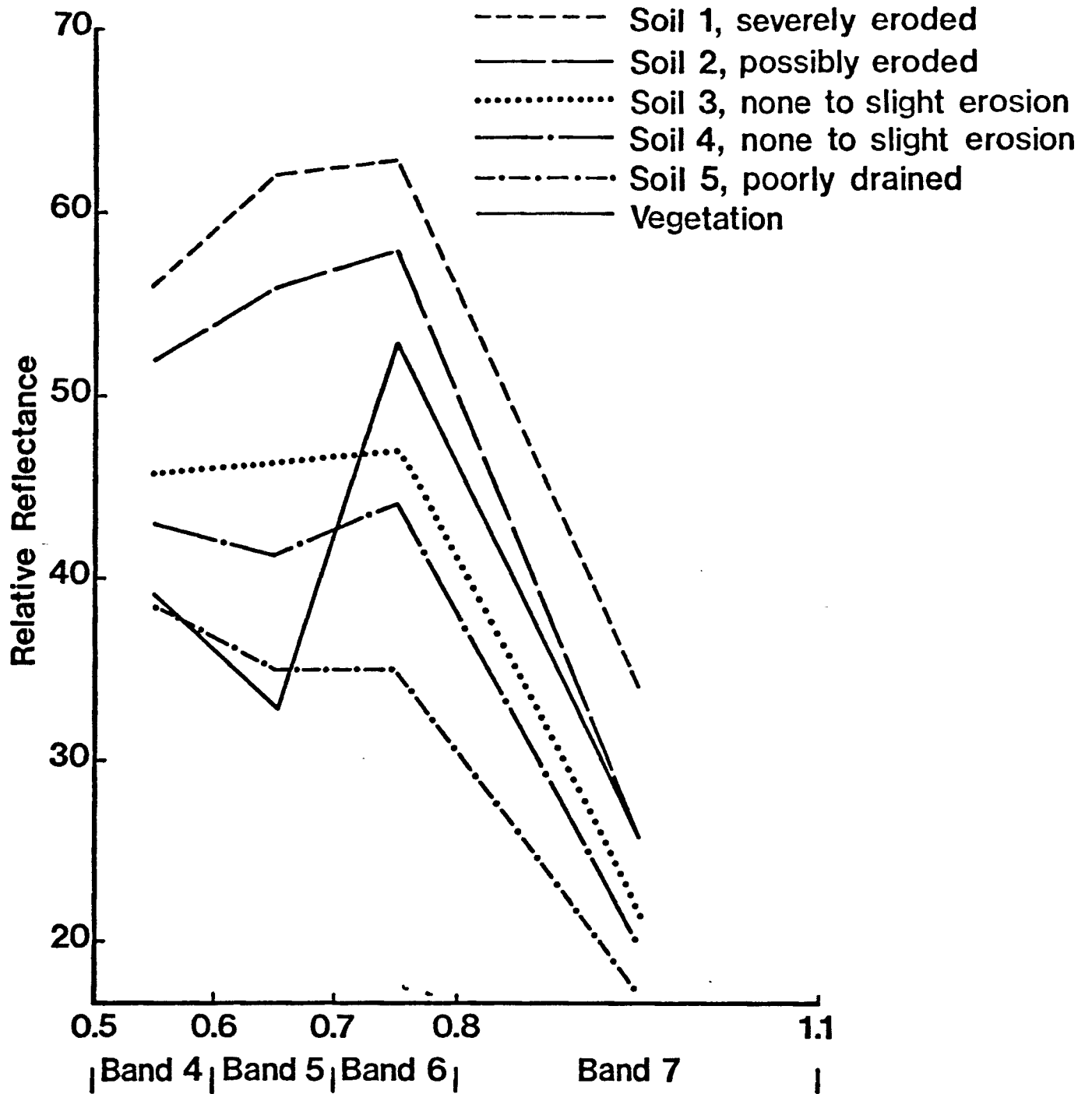


Figure 9. Spectral curves derived from Landsat multispectral data delineating severely eroded soils from other classes.



Key

White - Wheat  
Black - Fallow Land

Gray - Permanent Pasture

Figure 10. Classification of Greeley County, Kansas from Landsat MSS data collected 19 June 1973.

Table 1 Specifications of Sensors on Landsats 1, 2 and 3

<u>Landsat</u>	<u>Instrument</u>	<u>Spectral Range</u> ( $\mu\text{m}$ )	<u>Description</u>	<u>Spatial Resolution</u> (m)
1, 2	RBV	0.475 - 0.575	Blue-green	80
		0.580 - 0.680	Orange-red	80
		0.690 - 0.830	Near Infrared	80
3	RBV	0.505 - 0.750		38
1, 2, 3	MSS	0.5 - 0.6	Green	80
		0.6 - 0.7	Red	80
		0.7 - 0.8	Near Infrared	80
		0.8 - 1.1	Near Infrared	80
		10.4 - 12.6	Thermal Infrared	240



Table 2 Specifications of the Thematic Mapper of Landsat D

<u>Spectral Bands (<math>\mu\text{m}</math>)</u>	<u>Description</u>	<u>Spatial Resolution, IFOV (m)</u>
0.45 - 0.52	Visible Blue	30
0.52 - 0.60	Visible Green	30
0.63 - 0.69	Visible Red	30
0.76 - 0.90	Near Infrared	30
1.55 - 1.75	Middle Infrared	30
2.08 - 2.35	Middle Infrared	30
10.40 -12.50	Thermal Infrared	120