URBAN LAND-USE MAPPING BY MACHINE PROCESSING OF ERTS-1 MULTISPECTRAL DATA: A SAN FRANCISCO BAY AREA EXAMPLE

BY

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# The Laboratory for Applications of Remote Sensing

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## URBAN LAND-USE MAPPING BY MACHINE PROCESSING OF

#### ERTS-1 MULTISPECTRAL DATA:

### A SAN FRANCISCO BAY AREA EXAMPLE\*

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#### ABSTRACT

Results are promising of an experiment to map land use in an urban area by automatic digital processing of ERTS-1 data. Computer-maps of a large segment of the San Francisco-Oakland and San Jose Urbanized Areas have been produced at a scale of 1:24,000 using a segment of an ERTS-1 frame reformatted to correct skewness and scale. An area of some 6,500 square kilometers was also mapped at 1:48,000 (a one-fourth sample). For both scales, urban areas were separated from rural -- using a photo interpretation procedure -- to solve problems of the spectral similarity of functionally different land uses and land covers.

Classification was achieved by grouping twenty-eight spectral classes into eleven functional classes. Reliability was checked by comparing computer results to contemporaneous high-altitude color air photographs on a pixel-by-pixel basis. Performance results are high considering the grossness of the data and the complexity of the urban landscape.

#### INTRODUCTION

Discussed here are the results of attempts to create computer-produced urban land-use maps using multispectral scanner data from a satellite. The study is an outgrowth of research questions posed by individuals connected with LARS/Purdue, NASA, the Earth Resources Observation Program of the Department of the Interior, and the Geographic Applications Program of the U.S. Geological Survey.

Specific study objectives have been: (1) by LARS to test the applicability of the LARSYS pattern recognition software to urban land-use studies in an area where contemporaneous ground truth was available; and (2) by the Census Cities Project of the Geographic Applications Program (Wray, 1972) to attempt to utilize ERTS-1 data as a support or possible replacement for land-use mapping achieved through conventional air-photo interpretation. Further Geographic Applications Program goals are to utilize maximally the ERTS-1 data to: (1) produce print-out maps of large (1:24,000) scale; (2) aggregate digitized land-use data which may be used in conjunction with such reported ground-collected data as census reports and parcel ownership; and (3) monitor urban change on a regular basis.

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For the land-use mapping, the goal of this work has been to attempt to delineate with maximum accuracy as many functional land-use classes as possible given the existing limitations of data and processing techniques. The term functional must be emphasized as it pertains to usage of land, and not to naturally occurring land cover. Further, these functional land uses are selected to correspond as closely as possible to classes which are widely accepted and used by those in the planning and land management community. While the usual technique in mapping land use from air-photo interpretation is to infer function from the visible and associative characteristics of the imagery, inference of functional land use via automatic machine processing in this case must be accomplished from the only information available, namely the spectral data for each resolution element (pixel) telemetered to earth from ERTS-1. In short, spectral characteristics must be translated into meaningful, acceptable, land-use classes. A LARS paper (Hoffer, 1972) makes a similar point.

Workers facing this translation problem have proposed varying solutions. At LARS, a land-use map of Milwaukee County (Todd, 1973) was prepared with thirteen classes some of which were statements of broad urban patterns such as "inner city" and "suburban" while others were of such discrete land cover as grass and five classes of water. Work at Johnson Space Center (Erb et al, 1973) took a more general tack in defining six broad classes which combined spectral and functional characteristics. Included were "large buildings and building complexes with high-reflective roofs" and "open grass-covered fields with few trees."

A further consideration throughout this research effort has been to approximate as closely as possible (1) the land-use classes employed in the Geographic Applications Program's mapping of the San Francisco nine-county test site (Ellefsen and Peruzzi, 1972) and (2) the proposed system of the Geological Survey (Anderson et al, 1972). The attempt has thus been made to determine the limits of inferring land uses from spectral information alone. This approach was selected realizing that some urban planners (Grey, 1973) argue that intensive, accurate land-use determination from remote-sensed imagery of any sort is not possible and that the plotting of such discrete uses as retail, office, and many multi-family residential units are possible only with parcel-wise data secured from ground sources.

#### DATA PROCESSING

For this analysis, data were selected from the July 26, 1972 ERTS-1 frame (system corrected digital data from the multispectral scanner) which includes the heavily urbanized area on the east, south, and west sides of San Francisco Bay. The multispectral imagery for this area was first subjected to a geometrical preprocessing operation on the computer which (1) rotated the orientation to northsouth, (2) removed the skew due to the earth's rotation, and (3) rescaled the data so that computer line-printer maps would have a scale of 1:24,000. In the preprocessed data, each multispectral data point represented a rectangular area on the ground, 61.0 m by 76.2 m (.465 hectare).

The data were then analyzed using the pattern recognition techniques implemented in LARSYS, the remote sensing data processing system developed by LARS (Swain, 1972). In particular, an "unsupervised classification" approach was employed: cluster analysis was used to isolate spectrally distinct classes in the multispectral imagery and available ground truth data were then used to associate a ground-cover description with the resulting spectral classes. A systematic ten percent sample of multispectral data from an urban area in the vicinity of San Jose was subjected to the cluster analysis. A total of 5226 resolution elements were clustered into 30 spectrally different classes. Two of the resulting classes were later deleted because they contained too few points for computing second order statistics. A printout of the classes was then compared against ground truth, viz. contemporaneously taken high-altitude color infra-red air photographs (1:130,000 scale RC-10 diapositives flown by U-2 aircraft by NASA in support of GAP's Census Cities Project). The photographs were superimposed over the computer print-out with the aid of a Bausch and Lomb Zoom Transfer Scope. Precise determination of character of land use, on a pizel-by-pixel basis was thus possible. Results were tallied and then analyzed to determine the best grouping of the spectral classes into functional classes. A total of eight categories of urban land uses and three rural land uses were finally selected as presented in Table 1.

A further concern throughout the experiment has been to determine how well automatic processing of ERTS-1 data interacts with the scheme for the classification of land use based on remotely sensed data proposed by the Geological Survey (Anderson et al, 1972). That system contains levels of generalization I and II. More discrete level III uses, collapsible into the more general classes, may be developed by users. Comparison with the classification system developed in this experiment (see Table 2) demonstrates, at least for the studied urban sample, that machine processing of satellite scanner data is capable of a much finer classification than simply Level I. Level II is achieved in the delimitation of residential use, open space, and a combined commercial and industrial class. In addition, the very fact that we are dealing with spectral characteristics permits (even requires in order to maximize the available data) the classifying of parking lots and mobile home parks and differentiating between improved and unimproved open space and thus enters the realm of the discrete uses treated in Level III.

In short, the combination of the satellite-borne scanner and machine processing provides a different tool than either conventional air-photo interpretation or surface and statistical unit mapping. Each produces a somewhat similar but yet different product and each has advantages and disadvantages. While these have been documented for air photography and ground methods, the characteristics of machine-processed satellite data as applied to urban mapping are not well known and deserve presentation.

The advantages are:

High speed processing

Frequently obtained new data

Unbiased and uniformly repetitive classification

4. Production of print-out maps at a large map scale at relatively low cost (once the system becomes operational)

The inherent digitizing of land-use data retrievable in virtually any form or combination of forms

The disadvantages are:

The inability of the system to discriminate with consistent success between functionally dissimilar but spectrally similar land uses The impossibility of detecting parcel ownership

3. Generalization by resolution element: at 80 meter resolution the complexity of the urban landscape cannot be shown fully

Identifications dependent on vegetation vary seasonally

5. Uncontrollable incidence of cloud cover

Review of the above suggests that for many potential users, the satellite/machine-processing system has advantages which outweigh disadvantages and will be welcomed as a new, powerful tool in spatial analysis work. Other users may have to await refinements in the system which will surely come with subsequent developments in scanner and data processing capabilities.

#### TEST AREA

Computer maps were created at three scales. For the largest -- at full size (all pixels) for a map scale of approximately 1:24,000 -- a total area of about 1,125 square kilometers was mapped. Corporate units include Oakland, Alameda, San Leandro, Hayward, Fremont, San Jose, Santa Clara, Mountain View, Sunnyvale, Palo Alto, Menlo Park and several lesser suburban municipalities.

A larger area, some 6,500 square kilometers, was also mapped at a scale of 1:48,000. A one-fourth sample was achieved by instructing the computer to classify and print out alternating pixels per column and per line. Only one-fourth of the data were treated; they were not averaged.

A similar approach was followed in preparing a computer map at a scale of 1:72,000 for an even larger segment of the area of the nine San Francisco Bay Counties by using the device of a one-ninth sample (every third pixel by column and by line). Though skew and scale were uncorrected, the product served to demonstrate the possibility of mapping large urban and urban related areas; a total of approximately 21,000 square kilometers were mapped.

The test area has many general characteristics in common with rapidly growing cities throughout the United States. Development has been more horizontal than vertical with: large areas of new single-family residences built on converted agricultural land; several large clusters of new industrial complexes; shopping centers; and various institutions. The original core areas of the nuclei cities, from which growth spread in the past twenty years, remain as islands of old within the new, but many of these have been significantly altered. Connecting ribbons of commercial development are visible even to the unaided eye on the ERTS-1 imagery. Small patches of unbuilt-upon agricultural remnants remain as enclaves while exclaves of the expanding city are found in the rural-urban interface area.

#### CLASSIFICATION PROBLEMS

A key question to be examined at this stage in the experimentation with mapping urban land uses by automatic digital processing of satellite imagery is how reliably functional land-use classes can be derived from spectral data. The approach to an answer requires exploring in depth the spectral characteristics of urban features. Considering that many components of the urban scene are smaller than the 80 meter square pixel and that great spectral diversity occurs from place to place within an urban area, an examination of the structural components of each urban spectral signature is necessary. For example, residential land use is composed of such spectrally diverse features as asphalt streets, concrete drives and patios, shake roofs, varying levels of maturity of landscaping, "corner" grocery stores, churches, and schools. In addition, these vary regionally with different environmental conditions and local varieties of building and paving materials.

In another example, commercial-industrial use following its general tendency to be conducted in a specialized urban environment, yields a unique spectral signature because it is nearly always found in buildings or clusters of buildings with flat roofs with sizable parking and storage areas adjacent. The surfaces are spectrally quite distinct from a residential area with its pitched roofs, landscaping, and a full network of accompanying streets.

Where a confusion of symbols is seen on the print-out map, ground-truth examination reveals that such areas are indeed quite diverse and present classification problems even to the land-use mapper on the ground. Cases of incorrect identification are often simply cases of recent or on-going construction, areas which have not had sufficient time to weather into a more typical spectral signature.

A number of identification problems are common to all of the classes involving man-made cover of the land. Of greatest concern is that functional use is not consistently reflected in building shape (seen in photographs) or spectral characteristics (recorded by the scanner). The user of urban land-use data has a real need to distinguish between such diverse functions as retail, education, wholesaling, and transportation. When all these functions are found in spectrally similar settings, discrimination using spectral information alone is impossible. Attempts to determine reliable signatures for commercial versus industrial have proven inconclusive: while commercial establishments, such as along arterials and in shopping centers, usually have asphalt roof surfaces, industries exhibit bright and dull surfaces in about equal quantity.

An added problem is the differential weathering of all types of man-made surface materials. Old and new paving and roofing materials are spectrally distinct enough for the computer to classify them differently even though functional land use is the same throughout. The ability of the computer to make this distinction suggests a potential to differentiate newer from older developments, but this potential has not yet been exploited.

#### EXPERIMENT RESULTS

The experiment has so far produced several computer maps of varying scales. First, all of the major urban complexes of the San Francisco Bay Region (on the

single ERTS-1 frame employed) not covered by cloud have been mapped at 1:24,000. All of these lie within the segment of the frame corrected for skewness and scale and even though further adjustments are required, the registration of the computer map to the  $7\frac{1}{2}$  minute U.S.G.S. quadrangle map is remarkably accurate with perfect registration falling off only some one to two pixels over a distance of about 16 kilometers. Overlaying the computer maps over the quadrangle maps on a light table shows an immediate high correspondence with the quadrangle's limited landuse information. A temporary limit has been reached with the eight urban and three rural land-use classes selected. Isolating additional classes by determining the best sub-groupings from within those now comprising several spectral classes must await the application of new experimental techniques.

For all of these maps (plus the 1:48,000 one-fourth sample), it has been necessary to separate rural from urban uses in order to overcome the persistency of the classifier to throw into the same class certain rural (usually agricultural) land uses with certain urban ones (see Figure 1). A common confusion is caused when both urban residential land and cropland occupy the same spectral class. It is a matter of coincidence that the combination of ambient soil cover—especially if somewhat moist—and an immature crop, is spectrally similar (within a total of twenty-eight classes) to a single-family residential area with its asphalt streets, dark shake roofs and a given amount of landscaping. Fellow experimenters at LARS, Houston, and the University of Illinois have encountered the same problem.

The method employed to date to by-pass the problem requires the preliminary step of delimiting the urbanized area; subsequent groupings of urban land uses lie inside the boundary line and rural groupings are outside the line. A procedure was followed (Peruzzi, 1973) which adapted the Census Bureau's rules for Urbanized Area delimitation to a one-quarter kilometer grid system. The example (in Figure 1) uses a one kilometer cell generalization. Kilometer squares were given UTM addresses and the corresponding computer coordinates were then entered into the computer and delimited on the print-out. Two separate groupings of the twenty-eight cluster classes, one urban one rural, were then printed out. Manual cut-and-paste techniques were then used to make a single map; these functions will be performed by the computer in future work. The one kilometer unit of generalization is also a little too gross; refinement to the quarter kilometer will further improve the product.

The introduction of the kilometer grids into the system also provided a basis for aggregating land uses by a standard areal unit. Table 3 illustrates for a typical few kilometers and for the average of 250 square kilometers the percentages of each land use for the area around San Jose. The figure of 62.0 percent for residential uses compares favorably with the 63.4 percent for approximately the same area as measured by planimeter from air-photo interpreted uses in the work of the Geographic Applications Program's Census Cities Project.

Knowing the reliability of automatic machine classifying is of greatest importance at this early stage of the work. Testing is hindered, however, since a precise definition of reliability and the development of a method of measurement are not universally agreed upon. Others have measured accuracy quite generally on an area basis and against the ground-truth of published land-use maps. In this experiment, accuracy of classification figures are based on a comparison of the classified individual pixels to contemporaneous air photographs. The procedure employed was tedious and time-consuming but simple in its method. Upon superimposing the photo over the computer map on the Zoom Transfer Scope, the question was asked for each pixel (in a 13 square kilometer sample) if the real land use matched that given by the classifier. Score was kept and the results presented in Table 4.

Reasons for mis-classification were readily recognized in the process of checking. First, the grossness of the 80 meter resolution elements vis-a-vis the size of urban features causes considerable unavoidable error. A common occurrence is where a row of symbols of an adjacent use covers a linear feature such as a highway arterial. Being linear phenomena highways appear to the eye viewing a photograph to be wider than they actually are; a four-lane highway has only some 20 meters of asphalt or concrete surface. Part of a roadway's signature is median strip and shoulder and in cities a commercial arterial is visible on a constructed ERTS-1 image only because of the distinctive reflectivity of the

flat-roofed buildings facing it.

Other frequent mistakes include confusion between the bright dried grass of a vacant lot and a bright factory roof or the unusually heavy tree canopy of an older, well developed residential neighborhood and a naturally wooded open space area.

It is possible to view in Figures 2, 3, 4, and 5 the degree of visual correspondence between computer map and photograph. The general land-use patterns of Hayward are demonstrated in the first two illustrations while in Figures 4 and 5 it is possible to see on a pixel basis just where classification is correct and where it fails. In reliability checking with the Zoom Transfer Scope, where the view is of a similar scale, the operator begins to "think" like the computer classifier. More of this intensive work should lead to classification refinement.

Figure 6 presents the one-fourth sampling computer land-use map along with the RBV image and a point-line identification map. In addition to general physical features visible on the map, many broad urban patterns are also recognizable such as commercial land uses along arterials, central business districts, commercial-industrial concentrations, residential areas, and open spaces. Rural land uses outside the delimited urbanized area, in accordance with the simplistic classification employed, render an adequate representation of the grass and tree land-scape of the non-urbanized hills and valleys lying between San Francisco Bay and the Central Valley. To date, attention in this experiment has been directed mainly at solving urban use classification problems. Classification of rural land uses and land cover will be attempted later.

#### CONCLUSIONS

Results of the experiment to date demonstrate that producing land-use maps of a large scale by machine processing of ERTS-1 scanner data is feasible. By keeping land-use classes fairly broad, a remarkable level of accuracy is attained despite the relatively coarse resolution and the inherent complexities of manmade land cover.

Much follow-on work is required. One constant challenge lies in improving the reliability of land-use recognition and classification. A finer sorting of the spectral information is one road to follow. Another would be the development of algorithms using context to solve certain classification problems. These techniques -- somewhat analogous to identification by association procedures used in photo interpretation -- are required for the computer to make distinctions between such functionally different but spectrally similar land uses as factories and shopping centers.

A second major thrust must be made toward developing and fitting optimal classification schemes to the capabilities of scanner-produced data from Skylab, ERTS-B, and aircraft from varying altitudes. The larger map scale (and the smaller area of each resolution element) of aircraft-borne scanners may increase the reliability of use identification and lend insight to work with ERTS-1 data. The inclusion of a thermal band, as planned for ERTS-B, would add a useful variable.

An additional effort is required to utilize computer graphic techniques to improve the visual quality of the land-use maps to promote their greater usage. A tailoring of products could be achieved to accommodate the entire range of users from local to national.

Data aggregation could be expanded from the kilometer squares in this paper to include census tracts, corporate units, transportation zones, and any other areal groupings desired. The ability to aggregate land uses by such areas and to monitor change with great frequency holds enormous promise for such valuable measurements as intercensal population estimates.

Essential too is the study of a temporal series of ERTS-1 passes for the purpose of monitoring and detecting change. Many of the chores presently done by hand lend themselves to machine processing. Important products would be precise measurement of incremental growth of subdivision housing, commercial, and trans-

portation uses. Summary statements of the type of change from one use to another could be facilitated by machine processing. A further advantage would be the ability to detect the fairly small-area changes within the body of the old city at the large map scale (1:24,000) of the computer print-out.

The solution of several other practical problems could be furthered by the use of a scanner/computer-analysis system. A relatively simple task would be the frequent up-dating of the boundary of a city's urbanized area. Commercial applications are also possible in such common jobs as selecting optimum locations for stores, banks, and service stations. The location and measurement of open space, a matter of key general concern, would also be easily handled by such a system. Another practical problem which could be dealt with is the required measurement of land use, present and projected, as a basis for mass-transit planning. Also, careful work may yield a method for measuring housing quality.

In sum, the advantages of speed, data handling, relative low cost, and frequent synoptic monitoring could be of extreme value in helping to solve many land-use problems.

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Table 1. Functional Land-Use Classes Employed on Computer Maps

*****	Functional Land-Use	Spectral Classes Comprised
URBAN	Commercial-Industrial Mobile Homes Residential	1, 2, 3, 14 5 6, 9, 10, 13, 15, 16, 17, 18, 19, 20, 21 8, 22
RURAL	Parking Lots Unimproved Open Space (bare) Improved Open Space (irrigated) Unimproved Open Space (with trees) Water	11 12
	Grazing and Cropland	1, 2, 3, 5, 6, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23 24, 25, 26, 28, 29, 30
	Tree Covered Water	24, 25, 26, 28, 29, 30 27

Table 2. Comparison of a Land-Use Classification Derived from Automatic Machine Processing of ERTS-1 MSS Data with a U.S.G.S. Proposed Land-Use Classification System for Use with Remote Sensor Data

Machine-processed ERTS-1 Data	U.S.G Level I	.S. Proposed System Level II		
URBAN	rever r	rever ii		
Commercial-Industrial	O1. Urban and Built-up Land	O2. Commercial and Services; O3. Industrial; O4. Extractive; O6. Institutional; O7. Strip and Clustered Settlement; O8. Mixed		
Mobile Homes	01. Urban and Built-up Land	O1. Residential		
Residential	01. Urban and Built-up Land	O1. Residential		
Parking Lots	01. Urban and Built-up Land	05. Transportation		
Unimproved Open Space (bare)	O1. Urban and Built-up Land	09. Open and Other		
Improved Open Space (irrigated)	O1. Urban and Built-up Land	09. Open and Other		
Unimproved Open Space (with trees) Water		O1. Deciduous; O2. Evergreen; O3. Mixed O3. Chaparral O1. Streams and Waterways; O2. Lakes; O3. Reservoirs; O4. Bays and Estuaries; O5. Other		
RURAL Grazing and Cropland O2. Agricul- tural Land		O1. Cropland and Pasture; O2. Orchards, Groves, Bush Fruits, Vineyards, and Horticultural Areas; O3. Feeding Operations; O4. Other		
	03. Rangeland	Ol. Grass; O2. Savannas; O4. Desert Shrub		
	06. Nonforested Wetland	O1. Vegetated; O2. Bare		
	07. Barren Land	O1. Salt Flats; O2. Beaches; O3. Sand Other Than Beaches; O4. Exposed Rock; O5. Other		
Tree Covered Water	04. Forest Land 03. Rangeland 05. Water	O1. Deciduous; O2. Evergreen; O3. Mixed O3. Chaparral		
HAUGI	O). water	<ul><li>O1. Streams and Waterways; O2. Lakes;</li><li>O3. Reservoirs; O4. Bays and Estuaries;</li><li>O5. Other</li></ul>		

Table 3. Land-Use Aggregations by Kilometer Squares for a Segment of the San Jose, California Area

## Percentage of Land-Uses

UTM Grid Designation	Comm- Indus	Mobile Homes	Parking Lots	Res	Bare	Trees	Irrig	Water	Thresh- old
135-603 134-590 134-591 134-592 134-593 134-594 134-595	32.5 12.1 9.8 13.3 18.8 32.1 42.9	1.7 1.3 0.0 0.5 0.4 6.7 4.5	1.7 11.2 14.7 9.0 14.7 33.5 18.3	59.0 72.3 72.3 75.7 62.1 13.8 27.7	4.3 0.4 0.0 0.0 0.0 0.0 0.0 3.6	0.9 1.8 0.9 0.0 3.1 1.3 0.0	0.0 0.9 1.3 1.4 0.4 11.6	0.0 0.0 0.0 0.0 0.0 0.0	0.0 0.0 0.0 0.9 0.4 0.9 1.3
Average for 250 square kilometers	12.9	1.3	8.2	62.0	4.5	6.2	2.2	0.2	2.5

Table 4. Reliability Test of Land-Use Classification

Functional Land-Use Class	Percent Correct
Commercial-Industrial	. 84.6 . 77.8

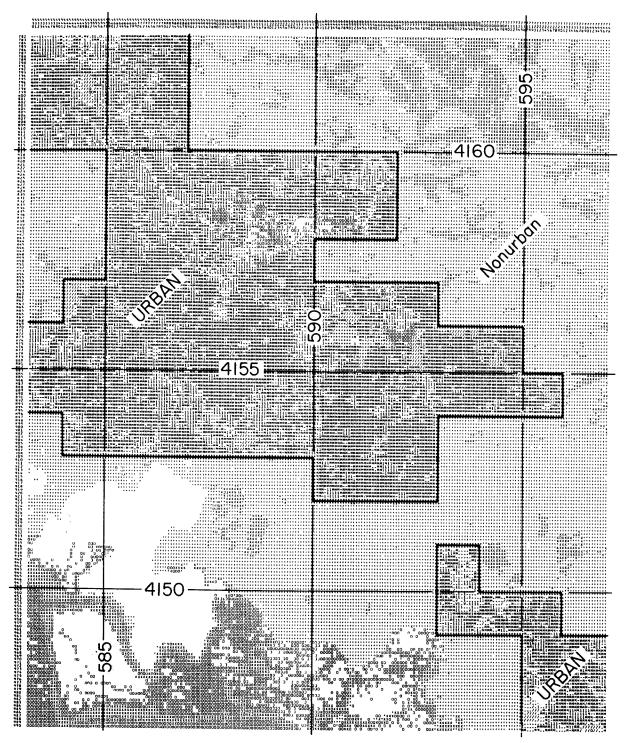


Figure 1. Computer-Classified Land-use Map of the Fremont Area, California. Map is derived from scanner digital tapes of ERTS-1 scene 1003-18175, 26 July 1972. It demonstrates use of separate classifications for Urban and Nonurban. Classification uses LARSYS pattern recognition algorithms, and was produced at Purdue University, Laboratory for Applications of Remote Sensing (LARS). Urban area is defined by one-kilometer UTM grid cell (zone 10) from USGS Census Cities ERTS experiment 1970 land-use map and NASA aircraft photography. Land use areas are aggregated by class and kilometer grid cell. Each pixel represent 0.465 hectares (1.1 acre). Urban classes: Commercial-Industrial (1); Mobile Homes (V); Residential (M); Parking Lots (·); Unimproved Open Space, Bare (-); Unimproved Open Space, Trees (/); Improved Open Space-Irrigated (+); Water (0). Nonurban: Grazing and Cropland ('); Trees (X); Water (0). Large unclassified areas (blank) are salt evaporation ponds.

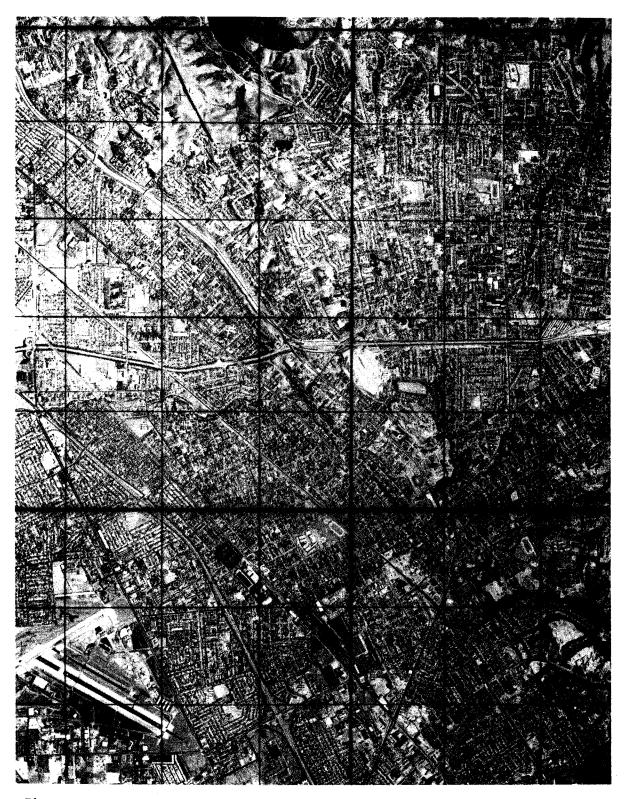


Figure 2. Black-and-white Photo of the Hayward Area, California. Photo by NASA for USGS Census Cities ERTS experiment: Zeiss camera, 12 in. focal length; 1: 50,000 from RB-57 at 50,000 feet; color infrared film (S0-117), filter D; frame 252, 15 May 1970. Approximate orientation of one-kilometer UTM grid (zone 10) indicates scale, and facilitates comparison with corresponding computer land-use map.

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Figure 3. Computer-classified Land-use of the Hayward Area, California. Map is from ERTS-1 scanner digital data for frame 1003-18175, 26 July 1972. Urban classes: Commercial-Industrial (1); Mobile Homes (V); Residential (M); Parking Lots (·); Unimproved open Space, Bare (-); Unimproved Open Space, Trees (/); Open Space-Irrigated (+); Water (0). Approximate orientation of one-kilometer UTM grid (zone 10) indicates scale and direction, and facilitates comparison with corresponding air photo and other ground truth.

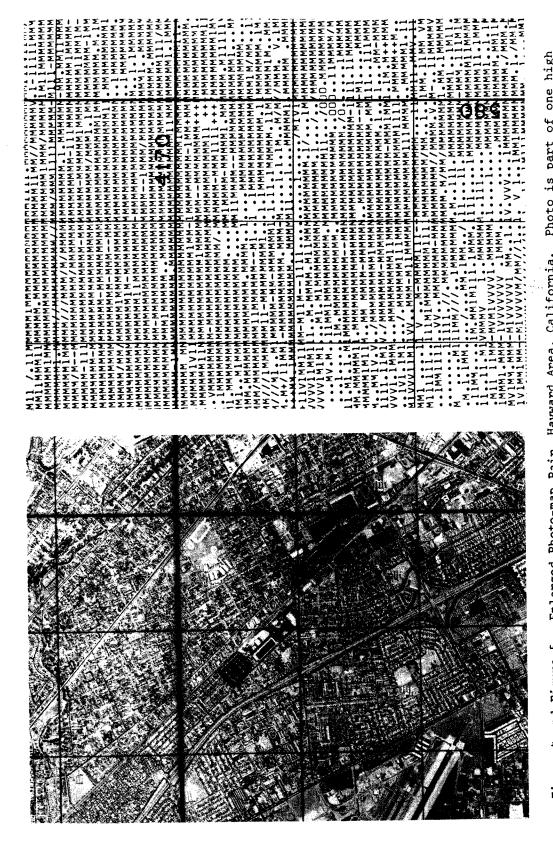


Figure 4 and Figure 5. Enlarged Photo-map Pair, Hayward Area, California. Photo is part of one high altitude color infrared scene acquired at 1:50,000, 15 May 1970, by NASA for USGS Census Cities ERTS experiment. Map is by Purdue/LARS and is based on ERTS-1 scanner digital data for frame 1003-18175, 26 July 1972. Scan lines have been reformatted so that they are nearly parallel to east-west UTM one-kilometer grid lines (zone 10). One printout symbol represents one ERTS scanner picture element (pixel), kilometer grid lines (zone 10). One printout symbol represents one ERTS scanner picture element (pixel), 0.465 hectare, or 1.1 acres. Urban classes: Commercial-Industrial (1); Mobile Homes (V); Residential (M); Parking Lots (.). Note correspondence between photo and map in mobile home area, lower left corner. Expansion of mobile home park (symbol V on ERTS computer map) is confirmed by NASA 1972 air photo.

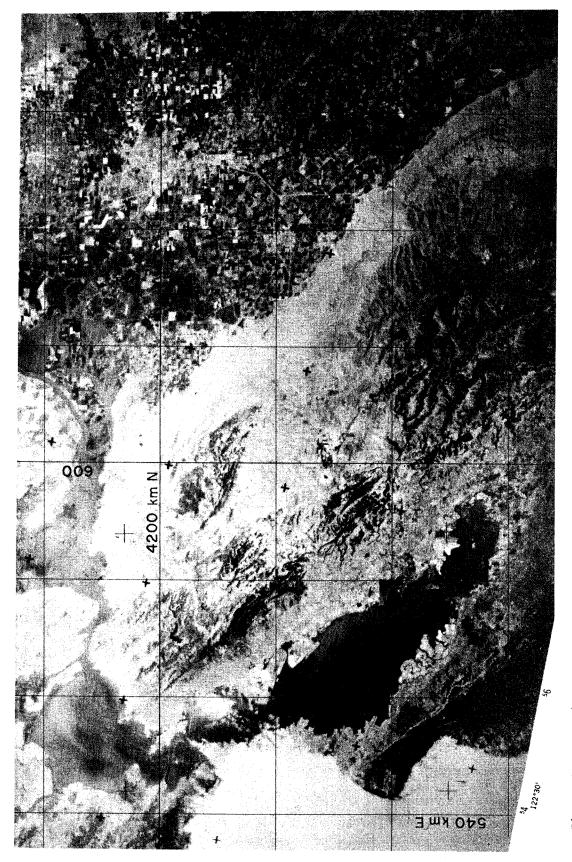


Figure 6a. Portion of ERTS-1 RBV View, San Francisco Bay Region. View is from Band 3 (red), frame 1003-18175, 26 July 1972, three days after launch. This is same view for which digital data from the fourchannel multispectral scanner (MSS) are used to classify land use by computer-aided techniques (Figure 6b). Twenty-kilometer UTM grid (zone 10) and thirty-minute geographic grid fitted by U.S. Geological Survey. Position of grids meet Federal map accuracy standards. Fog and San Francisco Bay at left (west); San Joaquin Valley and large irrigated fields at right; Hayward and Freemont areas at southeast side of San Francisco Bay.

