# Land Management Program 

THE CAUCA PROJECT


## THE CAUCA PROJECT (Land Management Program)

The study area for this project is located in the Cauca Department, which of Colombia. It lies approximately 50 km south of the city of Cali, and is centered on the geographic cordinates ( $76^{\circ}$ $30^{\prime} \mathrm{W}, 2^{\circ} 40^{\prime} \mathrm{N}$ ). The total area of interest extends to approximately ?? $\mathrm{km}^{2}$, and within this region altitudes range from ?? m to ?? m above sea level.

Climate: Mean Annual Temp/Rainfall/Seasons
Vegetation: Natural vagetation would be....
Now replaced by...
\{note spatial scale and complexity of land cover/use\}
The local economy is based on agriculture (subsistence + cash crops) Occupation of hill tops, with lines of communication running along crests Farming extending down the slopes towards the valley bottoms.

Although the topography is complex when viewed in detail, it has the general strucuture of a deeply dissected but gently sloping (which direction, $\mathrm{S}->\mathrm{N}$ ?) plateau. Slope gradients are typically high(??) and fall within the range of $0^{\circ}$ up to ?? ${ }^{\circ}$. Slope orientations vary widely and overall are approximately isotropic. Drainage is achieved via a number of rivers flowing approximately parallel from SE to NW. These are all tributaries of the Ovejas River which discharges into the Cauca River on the western limits of the study area. The study area is largely defined by the Ovejas watershed. However, within this watershed the tributaries to the Ovejas define a number of micro-watersheds, and amongst these the Rio Cabuyal has been selected for detailed study. Covering an area of ?? hectares, this micro-watershed is believed to be representative of the topography, agriculture and natural vegetion of the whole entire region.

## Aims:

The first aim in this project is simply to derive land cover information for the study area.
Information about basic land cover has been absent to date, but it is essential for a wide range of modelling procedures. These are concerned with: hydrology and water balance (e.g. Topog_IRM), the study of nutrient movements, a better understanding of soil development, and assessment of sediment movements and erosion. The ultimate aim of this modelling will be to assist in the identification and development of sound agricultural practices. These can be defined as those systems that provide high productivity that is sustainable and which does not cause environmental degradation.

A map of current land cover will also act as a bench mark against which future land cover changes can be assessed, and will allow strategic planning to be undertaken. Although not an immediate concern, it is hoped that archival data in the form of Landsat MSS imagery and air photography will eventually allow historical land cover changes to be assessed, perhaps as far back as 1940. There are, however, many technical difficulties to be overcome if this objective is eventually to be met. As an example, there will be a problem in matching the thematic information obtainable from the Landsat TM and SPOT HRV scanners used in the present study, to that which can derived from the MSS scanner with its lower spatial and spectral resolution.

The same problem of equating information between sensors of differing capabilities also arises in respect to another future aim of the GIS Land Management Group. Namely that, once it is estabished for the Cabuyal and Ovejas catchments, the group will be interested in extending land cover mapping to larger scales within Colombia and to other regions of the Tropics. On both logistical (ie data volume) and cost grounds it is certain that relatively coarse resolution imagery such as that provided by the MSS will need to be utilised. The accuracy and usability of MSS derived products need to be quantified and assessed in respect to higher resolution couterparts. Again, there are many potential difficulties and problems which lie ahead when attempting to equate information across sensors and scales (see, for example, Moody and Woodcock, 1994). However, a methodology in which MSS data is used to provide a general inventory, with TM/HRV and air photographs employed for detailed study in specific environments, is very appealing.

Objectives:
At a meeting held with the Hillsides Group, a hierachical land cover classification scheme was developed. Each level in this scheme would provide data of value and interest to the group. As thematic discrimination increases within these levels, the technical difficulties in achieving the product become more complex. The basic plan was to 'start at the easy end' and work towards the more ambitious products.

| (I) | (II) | (III) |
| :--- | :--- | :--- |
| LOW... | erosion... <br> fallow... | "infrastructure" <br> "natural processes" <br> fallow |
|  | arable... | other <br> casava <br> modern coffe <br> traditional coffee |
| MEDIUM... | pasture... | good quality <br> poor quality |
| HIGH... | woodland... | primary forestry <br> poor quality secondary <br> good quality secondary <br> managed reforestation <br> bamboo |

The main objectives were quite simply;
(1) Derive a classification at Level (I)
(2) Derive a classification at Level (II)
(3) Derive a classification at Level (III)
\{definition of detailed study area\}
\{details of data and imagery available\}
\{details of hardware/software\}
\{details of pre-processing to establish the basic database\}

## Image Enhancement and Information Extraction

There are a great many possibilities for enhancing the visual appearance of the images, and for transforming the radiometric data into more meaningful information. These include:
Contrast EnhancementsDecorrelation Stretch
Spatial Enhancement (e.g. spatial domain filtering)
Spectral Merges (SPOT Pan./Landsat TM/SPOT XS)
Band Ratios
DEM based topographic corrections
Principal Components Analysis
Other empirical transformations (e.g. Tasseled Cap)
Computation of spatial properties (edge density, texture etc)
Classification
where a number of alternatives arise...

- Unsupervised vs Supervised
-use of spectral bands only?
-include spatial informaton?
-include DEM information?
- hard classification?
- soft classification
maximum likelihood vectors
2nd most likely classes
spectral unmixing
fuzzy classes
- use post-classification filtering?
- use post-classification knowledge-based adjustments?
- use post-classification contraint-based adjustments?
- method of classification accuracy assessment

The principal purpose of all these procedures are to:

- Improve the visual display and aid manual image interpretation.
- Reduce uneven scene illumination due to topographic orientation, and negate its impact on the spectral signatures of land covers.
- Suppress common information between bands, and enhance spectral differences.
- Exploratory data analysis and dimensionality reduction
- Transformation of radiometric data to environmental information e.g. LAI, soil wetness, vegetation stress, etc.
- Transform radiometric vectors into nominal information classes or probabalistic functions.

Such a wide range of possibilites creates a certain dilema when plaaning a project. Which of the many possible avenues of image enhancement, transformations, classification schemes, additional datasets, etc should be followed.

In this particular instance there were two over-riding factors which shaped the direction of the work undertaken:
(a) the complex topographic environment, with steep slopes of all orientations
(b) the highly complex land use patterns, which are in themselves largely a result of the complex topographic environment. Individual farms are typically only ?? Ha in extent, and land use or land cover can change over a distance of ?? m. However, the diversity of arable crops is quite remarkable, although there is a certain degree of transience in the exact composition, driven by the market forces of cash cropping.

A key problem to be dealt with in this region is the potentially adverse influence of the complex topography on spectral signatures. Any attempt at a classification of land cover is likely to be influenced by topographic orientation in addition to the spectral signatures generated by differing land cover classes.

Topographic effects within the major valleys are clearly visible in the images derived from all three sensors. Furthermore, although they may not be so immediately apparent to the eye, it must be assumed that spectral signatures are equally influenced on a local pixel-by-pixel scale by the rapid changes in slope gradient and aspect over relatively short distances.

As a quick test of this assumption an unsupervised classification (ISODATA) was performed on the TM imagery and the correspondance of the spectral clusters with the slpoe and aspect maps derived from the DEM. Even by eye, it was immediately apparent that topography in isolation was having a significant influence on spectral responses.
\{\{chi-square of aspect/spectral class - being close to equator approx even number of days of sun $\mathrm{N} / \mathrm{S}$, and even hours $E / W$, so no reason to suppose topography should have a strong influence on vegetation composition\} use matrix, then take a random sample, then build matrix and put into
minitab\}
One solution to this problem is to treat the spectal signatures that are generated by a given land cover occuring on the 'dark' and 'light' sides of a valley as separate signatures during the classification process, and then to collapse this distiction in a post-classification stage.

This is rather crude since the distribution of slope aspects in the region is by no means bimodal, and thus a two-fold division of dark-light will incorporate considerable within-class-variability.

## Band Ratios

The 'classic' approach to solving this problem is to compute the ratio values between spectral bands. There are a number of reported examples where ratios have been found to be an effective solution (e.g. Hoben and Justice (1981), **).

However, excluding the thermal band there are 15 possible ratio combinations within the TM imagery. Thus, already, while potentially solving one problem this method can create another, namely which ratios to use?

| 1 | $1: 2$ | 2 | $1: 3$ | 3 | $1: 4$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| 4 | $1: 5$ | 5 | $1: 7$ | $6^{*}$ | $2: 3$ |
| 7 | $2: 4$ | 8 | $2: 5$ | 9 | $2: 7$ |
| $10^{*}$ | $3: 4$ | $11^{*}$ | $3: 5$ | 12 | $3: 7$ |
| $13^{*}$ | $4: 5$ | $14^{*}$ | $4: 7$ | $15^{*}$ | $5: 7$ |

So some extent arbitrarily, but with some respect of the current literature and an understanding of spectral signatures, the ratios indicated by * were computed.

Initial visual interpretation of the ratio data sets.
(4:3, 5:3) moderate topographic suppression
4:3 shows vegetation quantity
5:3 shows the clearest bare soil differentiation, but very little separation between vegetaiton types
(5:4, 7:4, 7:5) very good topographic suppresion
5:4 looks as if it provides good vegetation discrimination
7:4 similar in appearence to $5: 4$
7:5 same again, but a higher noise component
(3:2) moderate to poor topographic suppression
3:2 is generally noisy, unsure of information content
Many of the derived ratio images were, as might be expected, visually similar. To gain some measure of the degree of similarity a correlation matrix was computed between the ratios identified above and is presented in Table 1. This illustrates that the correlation relationships fall broadly into three categories:
(i) Very high
$(5: 4) \&(7: 4) \quad 0.96$
$(4: 3) \&(3: 2)$
0.81
$(7: 5) \&(7: 4) \quad 0.89$
(ii) High
...the rest...
(ii) Low

| $(5: 4) \&(5: 3)$ | 0.10 |
| :--- | :--- |
| $(7: 4) \&(5: 3)$ | 0.19 |
| $(7: 5) \&(5: 3)$ | 0.31 |
| $(3: 2) \&(5: 3)$ | 0.39 |

Table 1: Correlation Table for Band Ratio Images

| Ratio | $5: 4$ | $5: 3$ | $4: 3$ | $7: 5$ | $7: 4$ | $3: 2$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $5: 4$ | 1.0000 | -0.1016 | -0.7895 | 0.7769 | 0.9600 | 0.7749 |
| $5: 3$ |  | 1.0000 | 0.6175 | -0.3157 | -0.1930 | -0.3971 |
| $4: 3$ |  |  | 1.0000 | -0.7601 | -0.7843 | -0.8116 |
| $7: 5$ |  |  |  | 1.0000 | 0.8904 | 0.7409 |
| $7: 4$ |  |  |  |  | 1.0000 | 0.7743 |
| $3: 2$ |  |  |  |  |  | 1.0000 |

\{spectral distances and band combinations\}

## PCA Transformation

PCA is a linear combination that provides a rotation of the measurement axes in the feature space such that whilst remaining orthogonal they are aligned along the directions of greatest variance. Due to its derivation from the covariance/correlation between spectral bands the properties of a PCA is scene-dependant. It has been used for dimensional reduction and data compression as well as for the exploratory analysis of data structure. Higher components can often reveal subtle information not seen in the original spectral bands. There are numerous examples of its use in RS applications (eg ****). It was applied in this case for several reasons..
(1) To help understand the information content within the TM image
(2) To investigate the potential of using the PCA components in place of the original spectral bands
(3) To assess its ability to reduce dimensionality - which may become important if additional information layers were to be used in a maximum likelihood classification.

The correlation table for the TM bands (Table 2) shows the high interband correlations typical for imagery derived from this sensor. Particularly highly correlations occur between the visible bands, and also between Band 5 \& Band 7.

Table 2: Correlation Table for the Landsat TM imagery

| Band | 1 | 2 | 3 | 4 | 5 | 7 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 1.0000 | 0.9515 | 0.9197 | 0.2907 | 0.6202 | 0.6978 |
| 2 |  | 1.0000 | 0.9568 | 0.3780 | 0.7073 | 0.7568 |
| 3 |  |  | 1.0000 | 0.2438 | 0.7505 | 0.8354 |
| 4 |  |  |  | 1.0000 | 0.4670 | 0.2287 |
| 5 |  |  |  |  | 1.0000 | 0.9265 |
| 7 |  |  |  |  |  | 1.0000 |

The eigenvalues, when re-expressed as a cumulative percentage, demonstrate the potential value of a PCA transformation on this data set in reducing dimensionality with minimal loss of information. As shown in Table 3, the first three PCA bands carry over $98 \%$ of the total scene variance.

Table 3: Eigenvalues from PCA transformation of TM data

| Component | Eigenvalue | \%variance | \%cumulative |
| ---: | ---: | ---: | ---: |
| I | 651.73 | 70.87 | 70.87 |
| II | 168.64 | 18.34 | 89.22 |
| III | 87.28 | 9.49 | 98.71 |
| IV | 5.90 | 0.64 | 99.35 |
| V | 4.64 | 0.50 | 99.86 |
| VI | 1.32 | 0.14 | 100.00 |

Table 4: Component Loadings

|  | I | II | III | IV |
| :---: | :---: | :---: | :---: | :---: |
| 1 | 0.3403 | -0.2084 | -0.6360 | 0.5968 |
| 2 | 0.2321 | -0.0919 | -0.3294 | -0.1926 |
| 3 | 0.3255 | -0.2386 | -0.3294 | -0.6049 |
| 4 | 0.3020 | 0.9134 | -0.1968 | -0.1278 |
| 5 | 0.7045 | -0.0251 | 0.5476 | 0.3011 |
| 6 | 0.3696 | -0.2369 | 0.1998 | -0.3658 |

The component loadings are shown in Table 4. These show that PCA-1 is an approximately evenly weighted average of all spectral bands. PCA-2 is dominated by the information carried in TM Band 4, and PCA-3 picks up most information from TM Bands $1 \& 5$.

In an initial visual interpretation of PCA images the following comments were made:
PCA-1 Overall luminance across the sampled spectrum, or "brightness" Topographic detail and shadowing is very clear

PCA-2 Vegetation "lushness". This appears to match closely both the Tasseled Cap "greenness" and the 4:3 ratio (check via correls)

PCA-3 Separates valley bottoms and hill tops. Both vegetation and topo info present Appears to be an inverse of the Tasseled Cap "wetness" Thus it could be termed "dryness"?

PCA-4+ Were dominated by scanner and background noise and thus disregarded. They account for $<2 \%$ of total variance.

Since the topographic effect appears to be concentrated in PCA-1, one possibility is to "discard" this band and to use the remaining PCA bands in a classification scheme, thereby removing the topographic effect. However, with the loss of $70 \%$ of total scene variance, it is unlikely that this would be provide a good solution. Nevertheless, the fact that hyperspectral "intensity" is concentrated in PCA-1 is beneficial. It allows the possiblity of "correcting" this information in isolation from the other thematic information.

## Tasseled Cap Transformation

This procedure is based on a similar translation of the measurement axes in feature space. It is designed for vegetation studies, and the measurement axes have been rotated in such a fashion as to record information relating to dominant environmental influences: (Axis $1=$ soil brightness, Axis $2=$ vegetation biomass, Axis $3=$ soil and canopy moisture).

Unlike PCA, the rotation of the axes are not determined by scene variance but are a fixed translation determined by empirical observation. Originally defined for MSS data, then extended to TM (refs to Crist). For TM data Christ et al. defined four principal information carriers, termed "brightness", "greenness", "wetness", and "haze". One potentially big advantage of the tasseled cap is for crossscene comparisons, but the degree to which the interpretations described above match a particular scene under study is somewhat arguable. Since a Tasseled Cap is defined for MSS data also, it may be useful to compare the results of a TM and MSS based Tasseled Cap transformation for the same study region (a future project perhaps). If these prove compatible it would provide a 'link' between TM and MSS data to allow transitions across scale.

The Tasseled Cap and PCA transformations displayed very similar information. The visual similarity is borne out by the correlation matrix shown in Table 5. Clearly, the Tasseled Cap and PCA are to a large degree measuring the same information.

$$
\begin{array}{ll}
\text { TS1 \& PCA } 1 & (0.98) \\
\text { TS2 \& PCA } 2 & (0.95) \\
\text { TS3 \& PCA } 3 & (0.64)
\end{array}
$$

This implies that the Tasseled Cap transformation is effectively capturing the dominant information content of the 6 TM bands. A distinct advantage of the Tasseled Cap over the PCA transformation is that the consistency of the axes rotation allows cross-scene comparisons, something not possible
with PCA.
Table 5: Correlation Table for PCA (Ratio 4:3) and the Tasseled Cap

| Band | PC1 | PC2 | PC3 | TC1 | TC2 | TC3 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| PC1 | 1.0000 | 0.0000 | 0.0000 | 0.9818 | -0.2337 | -0.6783 |
| PC2 |  | 1.0000 | 0.0000 | 0.1506 | 0.9459 | 0.3553 |
| PC3 |  |  | 1.0000 | -0.1081 | 0.2192 | -0.6384 |
| TC1 |  |  |  | 1.0000 | -0.1118 | -0.5424 |
| TC2 |  |  |  |  | 1.0000 | 0.3516 |
| TC3 |  |  |  |  |  | 1.0000 |

However, in both the Tasseled Cap and the PCA images a noticable residual topographic effect is present in the non-"brightness" bands. The band ratio images appeared to be more successful in this respect.

## Backwards Radiance Correction and Spectral Merging

The literature suggests that spectral merging of SPOT PAN and TM is profitable (e.g. Carper et. al., 1990; Chavez et. al., 1991; Cliche G et. al., 1985; Harris et. al., 1990; Shettigara, 1992; Welch and Ehlers, 1987). This can be achieved by an RGB-to-IHS transform, replacing the intensity layer with the panchromatic data, and then reversing the process using IHS-to-RGB. Alternatively, it is possible to perform a PCA transformation, relace the Component-I image with the high resolution (panchromatic) band and then apply the inverse PCA transformation. (note: best to use PCs based on correlation matrix, not the covariance matrix).

The aim of this process is to combine the thematic "richness" of the TM imagery with the spatial detail of the SPOT HRV panchromatic data. Clearly, this is a tempting choice in this application where land cover is so spatially variable. Higher spatiall resolution is a distinct advantage in providing an accurate land cover map. However, to use this procedure would still leave the topographic effect to be dealt with. Taking ratios between bands after performing a spectral merge is becoming rather "messy" and convoluted. Furthermore, the effect of the ratio operation will be to largely suppress the information that has just been supplied by incorporating the SPOT panchromatic band anyway (\#i think\#).

The literature also suggests that topographic correction is more effective if based on a backwards radiance correction transformation (BRCT) rather than relying on band ratios (Civco, 1989; Conese et. al., 1993; Colby, 1991; Justic et. al., 1981; Naugle and Lashlee, 1992). This, of course, has greater demands in the sense that it requires a suitable DEM for the region (i.e. at an appropriate accuracy and scale). However, the use of additional information should enable

Thus, to negate the effects of topographic orientation and achieve as high a level of spatial detail as possible it was decided to attempt to combine these two procedures.
le:

1) Generate a DEM from either... air photo stereo imagery
(or the SPOT PAN overlap)
(or map contours)
2) Use the DEM to perform a BRCT using a Lambertian model, and the SPOT Panchromatic band.
3) Apply a spectral merge between the TM imagery and the corrected Panchromatic scene (using the PCA substitution method)
4) Supervised classifications would then be derived from - the original TM data

- the band ratio images
- the spectral merge with BRCT-correction

5) Using ground-truth data from air-photo interpretations - evaluate the classification accuracy - determine success of various approaches
(Discuss, especially in the light of the land cover and topographic characteristics of the Cauca region)

The utlimate test of all this, ie which is the "Best" can only come from an evaluation of the usefulness of the end product. In other words, how well the product allows the questions of the Land Management group to be answered, by providing land cover data at the most approriate compromise of spatial scale and thematic accuracy. Expert judgement will be needed in order to tell us this.

NOTES
Note| only a Lambertian model - may try Minnaert Constant to improve on this Note the correction does not identify topographic shadows - needs viewshed Note| the problem of TM sensor striping - should have used a de-striping method first such as Crippen (1989).
Note | need for a 'quality' DEM to derive accurate slopes
1:25000 contours interpolated to 10 m is just not good enough

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## Data Processing Operations

- The first task was to extract a subset of the full-scene images. Although the region to be rectified can be specified during the rectification process it is still quicker to display and move around a smaller sub-scene whilst collecting the GCP information.

A suitable sub-scene was identified using the Inquire Cursor feature whilst viewing the fullscene and identifying the file coordinates of a rectangular region that will encompass the area of interest.

The region selected must have reasonably generous margins around the actual study area to allow for the geometric rotation to be accommodated without loosing data near to the study area margins. Keep a record of the coordinates of the subset.

- Before rectification via Ground Control Points two things needed to be done.

1) any existing map projection information from the subscene image was remove (i.e. 'deleted') using Image Info
2) the image was edge sharpened to assist in searching for and placing GCPs. This can be done using a high-pass convolution filter, or alternatively with the "crisp" process available in IMAGINE.

If using a convolution filter specify your own coefficients, as the defaults tend to be a bit harsh. Good results are normally achieved from

| -1 | -1 | -1 |  | -1 | -1 | -1 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| -1 | 14 | -1 | or | -1 | 16 | -1 |
| -1 | -1 | -1 |  | -1 | -1 | -1 |

A sharpened image is only needed temporarily whilst locating GCPs - so only the 3 bands that are going to be used within the viewer (say 4,5,3 for a TM image) need to be processed and stored.

- The ground control points are located and entered using the GCP editor.

Ideally some 40 or so GCPs should be used for a top quality transformation, but in practice around 15 GCPs scattered evenly across the area of interest are usually sufficient. You can afford to drop some of these to get the reported RMS error down to an acceptable level, but the number should not drop below 8 .

An RMS error of 0.5 pixels is usually taken as a good level of accuracy, although this obviously depends somewhat on the specific project requirements. It is relatively pointless struggling for very high spatial precision if the data are to be merged or processed with other information at lower levels of accuracy. For example, if the TM data is rectified to an RMS accuracy of 0.5 (i.e. 15 m ), the SPOT Pan image need only achieve an RMS of 1.5 (i.e. 15 m ) to be comparable. High spatial precision is desirable, but in practice the trade off between achieving a perfect rectification and getting the work done should be kept in balance.

Take an ASCII copy of the GCP matrix once it has been set up; it is very time consuming if you have to start again for any reason. To do this, select the 5 columns of interest, then press the R.H. mouse button while positioned over the POINTS heading. Select EXPORT and supply a suitable file name.

Ònce the RMS error is at an acceptable level and the transformation coefficients established, a rectification report can be generated in a similar manner.

- Compute the transformation equation to perform the projection to a Transverse Mercator. Again, it is wise to save a copy of the transformation coefficients to an ASCII file.

A 1st order transformation is acceptable for high resolution, narrow swath satellite imagery. 2nd order transformations or higher should be avoided. Although they may provide a better fit to the GCPs, the warping involved in achieving this can actually causes a higher degree of spatial inaccuracy overall.

- Perform the rectification.

The spatial resolution of all the tiles was specified as a 10 m grid. Thus the TM and XS imagery is over-sampled. Cubic convolution interpolation was employed. This does alter radiometric values, and also led to complications with mosaicing (see later). However, the geometric integrity of the cubic convolution was believed to be important if cross-sensor spectral merging were to be performed later.

The final Cauca tiles are $5000 \times 5000$ pixels; the Cabuyal tiles are $2000 \times 1500$

## - SPOT rectification

Two SPOT scenes are needed to cover the Cauca Study Area. Each image was rectified to the same Transverse Mercator projection separately, and then a mosaic or stitching operation was used to join these two sections together.

Since the images were acquired consecutively on a common orbital pass there was no need to use any form of histogram matching operation. However, for some reason (probably a bug) the mosaic program tends to transform the image data via its respective LUT while stitching. This means that the LUT of each section of image needs to be removed, or set to have no effect, before the images are stitched.

## - SPOT mosaic operation

Although straightforward in principle, a complication arose due to the use of cubic convolution in the rectification stage. Basically, as the interpolation 'window' approaches and passes over the boundary of the unrectified image data the background zeros become incorporated into the interpolated output. This leads to a 'ramp' of decreasing values at the margins of the raw data files. When stitching two images, ERDAS will recognize a zero as the edge of the data. However, these non-zeros values are not recognised and so become incorporated into the output as an unwanted artifact.

The solution found to this was rather messy. For the panchromatic bands the raster
editor can be used to manually remove the effected edge pixels and replace them with zero values. This is moderately time consuming but quite effective. Unfortunately, for the XS data the process proved to be even more arduous. The problem here is that the raster editor only edits the red band image (for a reason not understood). Thus the solution became more complex...

- subset a single band from the XS image
- use the raster editor to set the affected edge values to zero
- use reclass on this layer to code all non-zeros to 1
- multiply the full XS data by resulting mask
- but this generates a float type data file
- thus, rescale or truncate the results back to an 8-bit scale.

This worked, but it was pretty time consuming.
Maybe there's a better way?

- Generating the ratio bands.

In theory the OPERATORS function under Image Interpreter could be used to generate the ratio band images. In practice it was found to be easier to use the SPATIAL MODELLER, first of all defining the graphical model shown below. This was then used to generate a script model (again shown below) which is easily edited to generate any band ratio combination. This model selects the relevant bands from the multispectral image, applies a subtraction based atmospheric correction, performs the division operation (catching divide by zero), and finally rescales the result back onto an 8 -bit range.

## Corrections:

Atmospheric corrections are usually recommended before performing a ratio operation. The histogram subtraction method is the simplest to apply, although the regression model approach should be easy enough to implement using the large reservoir or cloud shadows to calibrate the coefficients (see the Imagine manual for further details of these methods). In the current study the subtraction method was adopted.

From a study of the spectral values found within cloud shadows (for bands 1\&2) and the reservoir (for bands $3,4,5 \& 7$ ) the following correction factors were established (based on the integer average of 12 points)

1: $\quad$ subtract 46
2: subtract 14
3: subtract 11
4: $\quad$ subtract 5
5: subtract 2
7: subtract 1

## Stacking:

Once the individual band ratios had been computed a graphical model was used to combine the layers into a single multi-band file. The model and its associated script are given below.

## - Computing the Principal Components

There is a function in the Image Interpreter to perform this, but a graphical model was used
instead. PCA was performed on only 6 TM bands, the thermal band being excluded due to its discordant spatial resolution.

The Imagine Modelling Script was needed to derive an output of the Eigenvectors, Eigenvalues, and a correlation matrix. In fact, only a covariance matrix in computed by Imagine so a FORTRAN program was written to convert this into a correlation matrix. A FORTRAN program was also used to tidy up the output generated by Imagine.

## APPENDIX I

Remotely sensed imagery acquired for the Cauca region
Landsat MSS imagery:

| Date | 24-Dec-87 |
| :--- | :--- |
| Date | 01-Feb-76 |

Landsat TM imagery:
Path/Row 9/58

Date 07-Aug-89
SPOT XS imagery:
2 images
Date 29-Aug-87
SPOT PAN imagery:
2 images
Date 29-Aug-87
Azi: $\quad 71.7$
Elv: $\quad 67.1$
SPOT Stereo Pair:
ordered for Aug/1994...
Air Photography:
Dates: 1989 (43...)
1991 (148...)
Scale: 1:36000
Scanning Resolution: 20X10-6
Panchromatic film

## APPENDIX II

The dimensions of the study are were identified as indicated below:

## Cauca Study Site Limits:

Latitude/Longitude

$$
\left(76^{\circ} 45^{\prime} \mathrm{W}, 2^{\circ} 60^{\prime} \mathrm{N}\right)
$$



$$
\left(76^{\circ} 15^{\prime} \mathrm{W}, 2^{\circ} 35^{\prime} \mathrm{N}\right)
$$

Transverse Mercator
$\left(1035^{000}, 825^{000}\right)$


$$
\left(1085^{000}, 775^{000}\right)
$$

Centre Scene: $\quad\left(1060^{000}, 800^{000}\right)$
-
Cabuyal Subset Limits:
$\left(1052^{000}, 809^{000}\right)$


$$
\left(1067^{000}, 789^{000}\right)
$$

Centre Scene: $\quad\left(1059^{500}, 799^{000}\right)$
"Bogota" Transverse Mercator Projection Details
Projection:
Central Meridian
TM
Central Latitude
-77.082(02778)
False Northing
4.599(04722)

False Easting
1,000,000
Scale along meridian 1.0
1,000,000

## APPENDIX III

File structure and details of the imagery database

## DIRECTORIES

The processed imagery for the Cauca Study Area have been stored under two principle directories, /image 2 and /image with a file naming convention as follows:

1) Unrectified subsets taken from the full-scene data, and rectified $50 \times 50 \mathrm{~km}$ Cauca tiles, are stored under the directory /image $2 /$ mitch/.

Rectified $15 \times 20 \mathrm{~km}$ Cabuyal tiles, which are subsets of the Cacau tiles, are stored under the /image/cabuyal/ subdirectory, along with all subsequent data derived from processing and analysis.
2) Data obtained from different sensors are stored in separate subdirectories. All subdirectory names begin with either the letter $\mathbf{c}$, to donate the Cauca study area, or the letter $\mathbf{s}$, to donate the Cabuyal subset.

A subdirectory whose second letter is $\mathbf{u}$ indicates unrectified imagery, whilst $\mathbf{r}$ indicates rectified imagery.

The last 2 letters indicate the sensor from which the imagery are derived:
pn SPOT, panchromatic mode imagery
tm Landsat, Thematic Mapper imagery
xs SPOT, multispectral mode imagery
Thus, /image2/mitch/cupn/ indicates Cauca data, unrectified, SPOT panchromatic images.
|and, /image/cabuyal/crtm/ indicates Cabuyal data, rectified, Landsat TM images.

## FILE NAMES

These are constructed as follows:

1) The first 4 characters replicate the information outlined above. I.e., crtm, cuxs, srpn, etc.
2) In the case of the SPOT imagery, this is then followed by either $\mathbf{N}$ or $\mathbf{S}$ to identify whether the imagery is obtained from the northern or southern full-scene SPOT image respectively, or by NS to indicate the stitched imagery.
3) The remaining characters specify the date of image acquisition:

$$
\begin{array}{ll}
\text { Thus, } \begin{array}{l}
\text { cutm07au89.img } \\
\text { srpnNS29au87.img }
\end{array} \quad=\text { Cabuyal, rectified, stitched, SPOT Pan., 29th August } 1987 .
\end{array}
$$

4) Conventions for file endings is largely determined by ERDAS Imagine
.img image file
.cff Transformation coefficents file .gcc Rectification ground control points (GCPs)

However, the following will also be found...
.rep A rectification report file
.dat The GCP table (stored as an ASCII backup)
.crp The transformation coeeficients (stored as an ASCII backup)

## Cauca Study Area Tiles

```
image2/mitch/crpn:
crpnN29au87.img Cauca tile, northern Pan. image
crpnS29au87.img
crpnNS29au87.img
Cauca tile, southern Pan. image
Cauca tile, stitched Pan. image
image2/mitch/crtm:
crtm07au89.img
Cauca tile, TM image
image2/mitch/crxs:
crxsN29au87.img
crxsS29au87.img
crxsNS29au87.img
Cauca tile, northern XS image
Cauca tile, southern XS image
Cauca tile, stitched XS image
```


## image2/mitch/cupn:

```
* \(=\mathrm{N}\) or S to indicate northern and southern scenes respectively
```

```
cupn*29au87.img Subset of SPOT Pan. image
```

cupn*29au87.img Subset of SPOT Pan. image
cupn*29au87.gcc
cupn*29au87.gcc
cupn*29au87.dat
cupn*29au87.dat
cupn*29au87.cff
cupn*29au87.cff
cupn*29au87.crp
cupn*29au87.crp
cupn*29au87.rep
cupn*29au87.rep
/image2/mitch/cutm:
cutm07au89.img
cutm07au89.gcc
cutm07au89.dat
cutm07au89.cff
cutm07au89.crp
cutm07au89.rep
Rectification CCPs
Rectification CCPs
GCP table (ASCII backup)
GCP table (ASCII backup)
Transformation Coefficients File
Transformation Coefficients File
Transformation Coefficients (ASCII backup)
Transformation Coefficients (ASCII backup)
Rectification report
Rectification report
Subset of Landsat TM image
Rectification GCPs
GCP table (ASCII backup)
Transformation Coefficients File
Transformation Coefficients (ASCII backup)
Rectification report
image2/mitch/cuxs:

* $=\mathrm{N}$ or S to indicate northern and southern scenes respectively

| cuxs*29au87.img | Subset of SPOT XS image |
| :--- | :--- |
| cuxs*29au87.gcc | Rectification CCPs |
| cuxs*29au87.dat | GCP table (ASCII backup) |
| cuxs*29au87.cff | Transformation Coefficients File |
| cuxs*29au87.crp | Transformation Coefficients (ASCII backup) |
| cuxs*29au87.rep | Rectification report |

```

\section*{Cabuyal Study Area file structure}
```

Directories:
image/
cabuyal/
srpn/ Cabuyal, rectified, Pan. imagery
srtm/ Cabuyal, rectified, TM imagery
srxs/ Cabuyal, rectified, XS imagery
ap-samples/ Segments of 42.IMG resampled
dem/ DTM data and derivatives
mixed/ TM-SPOT merged data files
classified/ Classified imagery
crocabu/ Vector outline of catchment
rios/ Vector rivers coverage
vias/ Vector roads coverage
crocabu.evs Vector properties
rios.evs "
vias.evs "

```

Files:

\section*{image/cabuyal/srpn/}
srpnNS29au87.img Cabuyal tile, SPOT panchromatic image
image/cabuyal/srtm/ srtm07au89.img Cabuyal tile, Landsat TM image sttm-pca.img Cabuyal tile, TM PCA bands srtm6band.img Cabuyal tile, TM bands 1-5 \& 7 tasseled.ing Cabuyal tile, Tasseled Cap I-IV
image/cabuyal/srxs/ srxsNS29au87.img Cabuyal tile, SPOT XS image
image/cabuyal/ap-samples/
gt1EE.img Northern, most eastern air photo sample gt1E.img Northern, eastern air photo sample
gt1W.img Northern, western air photo sample
gt1WW.img Northern, most western air photo sample gt2EE.img Next south, most eastern air photo sample etc..
/image/cabuyal/classified/mixed/
stmpn-ap.img 1st attempt at supervised classification
ap-fins.sig
ap-sigs.sig
stmpn.img
stmpn.sig
```

/image/cabuyal/classified/tm/super/
tm-ratios.img Supervised, TM ratios
tm-ratio-super.sig Spectral sigs
/image/cabuyal/classified/tm/unsupr/
error.img "Error" image for level1.img
level1.img Unsupervised, TM ratios
level1.sig Spectral sigs. from ISODATA
tm-bands.img Unsupervised, Raw TM, 8 classes
tm-bands.sig Spectal sigs. from ISODATA
/image/cabuyal/dem/
cabuyal-dem.img DEM @10m from 1:25000 \& Hutchinson
mask.aoi Raster of Cabuyal catchment as an area of interest
brct-dem.img DEM sample
brct-asp.img Computed slopes from DEM sample
brct-slp.img Computed aspects from DEM sample
brct-img.img SPOT Pan. sample
brct-msk.img Defines area of interest
brct-out.img BRCT corrected SPOT Pan.
view*.img 3D views of Cabuyal, with TM/SPOT overlay
view*.prm Parameters of 3D view geometry
image/cabuyal/mixed/
stmpn.img Spectral merge of TM and Pan. (no correction)
stmxspn.img Spectral merge of TM/Pan./XS (no correction)
/image/cabuyal/srtm/ratio/
rat-class.img
ratios-old.img TM ratios - no correction
ratios.img TM ratios - with correction
ratios.sig

```

\section*{Other files:}
```

Map compositions are stored under
/users/mla/compose/
Graphical models are stored under
/users/mla/gmodels/

```

The directories and file names should be self-explanatory.

\section*{APPENDIX IV}

\section*{LANDSAT TM PRE-PROCESSING DETAILS}

Sub-scene coordinates from full-scene image:
\begin{tabular}{llll} 
ULX & 1750 & LRX & 4050 \\
ULY & 2400 & LRY & 4500
\end{tabular}

\section*{GCP Table:}
\begin{tabular}{llrrrr}
1 & GCP \#1 & 755.82 & 827.65 & 1049660 & 809310 \\
2 & GCP \#2 & 587.35 & 686.00 & 1045500 & 814000 \\
3 & GCP \#3 & 700.39 & 812.07 & 1048150 & 810000 \\
4 & GCP \(\# 4\) & 910.36 & 827.88 & 1054050 & 808710 \\
5 & GCP \(\# 5\) & 1074.06 & 872.81 & 1058460 & 806750 \\
6 & GCP & \(\# 6\) & 991.06 & 969.81 & 1055730 \\
7 & GCP \(\# 7\) & 708.88 & 1059.62 & 1047460 & 803380 \\
8 & GCP & \(\# 8\) & 1057.62 & 731.62 & 1058550 \\
9 & GCP & \(\# 9\) & 1217.62 & 809.12 & 1062780 \\
10 & GCP \(\# 10\) & 1356.81 & 1165.94 & 1065250 & 797990 \\
11 & GCP \(\# 11\) & 1368.81 & 1359.56 & 1064750 & 791800 \\
12 & GCP \(\# 12\) & 723.03 & 1437.84 & 1046260 & 792250
\end{tabular}

Transformation Coefficients:
1st order Transformation of cutm07au89.img
Imagine Tabular Report
\begin{tabular}{rrr} 
Row & \multicolumn{1}{c}{\(X^{\prime}\)} & \multicolumn{1}{c}{\(Y^{\prime}\)} \\
\hline 1 & -31629.312606 & 34160.599451 \\
2 & 0.034716 & -0.004961 \\
3 & -0.005010 & -0.034752
\end{tabular}

Rectification Report:
RMS error report
cutm07au89.img
Reported Total RMS error: 0.463281
Imagine Tabular Report
\begin{tabular}{|c|c|c|c|c|c|c|}
\hline Row & Point ID & \multicolumn{2}{|l|}{> X Residual} & \multicolumn{2}{|l|}{sidual RMS Error} & ution \\
\hline 1 & GCP \#1 & & -0.119 & 0.527 & 0.541 & 1.167 \\
\hline 2 & GCP \#2 & & 0.434 & -0.174 & 0.468 & 1.009 \\
\hline 3 & GCP \#3 & & -0.567 & -0.381 & 0.684 & 1.475 \\
\hline 5 & GCP \#5 & & -0.032 & 0.679 & 0.680 & 1.467 \\
\hline 6 & GCP \#6 & & 0.068 & -0.416 & 0.421 & 0.909 \\
\hline 9 & GCP \#9 & & 0.168 & -0.154 & 0.228 & 0.492 \\
\hline 10 & GCP \#10 & & -0.064 & -0.158 & 0.170 & 0.368 \\
\hline 12 & GCP \#12 & & 0.113 & 0.077 & 0.137 & 0.296 \\
\hline 13 & GCP \#13 & > & 0.000 & 0.000 & 0.000 & 0.000 \\
\hline Count & 9 & 9 & 9 & 9 & 9 & 9 \\
\hline Mean & N/A & 1 & -0.000 & -0.000 & 0.370 & 0.798 \\
\hline Minimum & N/A & / & -0.567 & -0.416 & 0.000 & 0.000 \\
\hline Maximum & N/A & / & 0.434 & 0.679 & 0.684 & 1.475 \\
\hline Stddev & N/A & / & 0.268 & 0.378 & 0.247 & 0.532 \\
\hline
\end{tabular}
```

File Name : crtm08au89.img
Last Modified : Sat Aug 6 12:12:17 1994
Number of Layers : 7
Layer Information:
Name : :Layer_1
Width : }500
Height : 5001
Type : Continuous
Block Width : 64
Block Height : 64
Pixel Depth :Unsigned 8-bit
Compression Type : None
Statistics :
Last Modified : Sat Aug 6 12:12:17 1994
Maximum Value : 255.000000
Minimum Value : 25.000000
Mean : 63.944597
Median : 57.000000
Mode : 56.000000
Standard Deviation : 32.713322
Projection Information :
Projection Zone : 0
Spheroid Name : WGS 84
Georeferenced to : Transverse Mercator
Map Information :
Upper Left center X : 1035000.000000
Upper Left center Y : 825000.000000
Lower Right center X: 1085000.000000
Lower Right center Y: 775000.000000
Pixel X size : 10.000000
Pixel Y size : 10.000000

```

\section*{SPOT XS PRE-PROCESSING DETAILS}

Northern Image
Sub-scene coordinates from full-scene image:
\begin{tabular}{lrrr} 
ULX & 0 & LRX & 3195 \\
ULY & 1700 & LRY & 3002
\end{tabular}

GCP Table:
\begin{tabular}{lllrrrr}
1 & GCP \#1 & 434.77 & 562.38 & 1046640 & 812280 \\
2 & GCP \#2 & 607.05 & 590.96 & 1049940 & 811220 \\
3 & GCP \#3 & 784.07 & 611.76 & 1053330 & 810260 \\
4 & GCP \(\# 4\) & 1271.82 & 798.83 & 1062500 & 805060 \\
5 & GCP \(\# 5\) & 1460.12 & 888.90 & 1065880 & 802730 \\
6 & GCP \(\# 6\) & 1389.62 & 1104.50 & 1068700 & 798830 \\
7 & GCP \(\# 7\) & 1560.67 & 1088.86 & 1067240 & 798500 \\
8 & GCP \#8 & 1454.79 & 1231.64 & 1064710 & 795950 \\
9 & GCP \#9 & 1054.68 & 1200.63 & 1056940 & 797840 \\
10 & GCP \(\# 10\) & 824.87 & 1217.95 & 1052740 & 798130 \\
11 & GCP \(\# 11\) & 488.67 & 1099.78 & 1046020 & 801490 \\
12 & GCP \(\# 12\) & 461.63 & 757.45 & 1046070 & 819480
\end{tabular}

\section*{Transformation Coefficients:}

1st order Transformation of cuxsN29au87.img
Imagine Tabular Report
\begin{tabular}{|c|c|c|}
\hline Row & X' & \(\mathrm{Y}^{\prime}\) \\
\hline 1 & -4467.476696 & 48465.167462 \\
\hline 2 & 0.049414 & -0.007504 \\
\hline 3 & -0.007775 & -0.049304 \\
\hline
\end{tabular}

Rectification Report:
RMS error report
cuxsN29au87.img
Reported Total RMS error: 0.7672
Imagine Tabular Report


\section*{Erdas Imagine File Information}

File Name : crxsNS29au87.img
Last Modified : Mon Aug 29 16:01:27 1994
Number of Layers : 3
Layer Information:
```

Name : :Layer_1
Width : 5001
Height : 5001
Type : Continuous
Block Width : 64
Block Height : 64
Pixel Depth :Unsigned 8-bit
Compression Type : None
Statistics :
Last Modified : Mon Aug 29 16:01:27 1994
Maximum Value : 255.000000
Minimum Value : 0.000000
Mean : 49.090765
Median : 44.000000
Mode : 0.000000
Standard Deviation : 37.807094
Projection Information :
Projection Zone : 0
Spheroid Name : WGS 84
Georeferenced to : Transverse Mercator
Map Information :
Upper Left center X : 1035020.000000
Upper Left center Y : 825000.000000
Lower Right center X: 1085020.000000
Lower Right center Y: 775000.000000
Pixel X size : 10.000000
Pixel Y size : 10.000000

```

\section*{Southern Image}

Full-scene image was processed:

\section*{GCP Table:}
\begin{tabular}{llrrrr}
1 & West Side Inter & 839.312 & 935.812 & 1046123.44 & 781922.81 \\
2 & Mid Intersect & 1316.187 & 862.187 & 1055771.56 & 781932.19 \\
3 & N-Mid Intersect & 1310.210 & 526.300 & 1056685.31 & 788596.56 \\
4 & W Side Main Rd & 1474.656 & 707.406 & 1059371.25 & 784526.25 \\
5 & W of Main Rd & 1322.531 & 129.531 & 1058136.25 & 796393.75 \\
6 & S of main Rd & 1500.350 & 567.076 & 1060317.81 & 787243.44 \\
7 & North of Fork & 1609.031 & 955.531 & 1061280.62 & 779230.62 \\
8 & South of Bend & 449.781 & 328.343 & 1040316.25 & 795081.25 \\
9 NW Intersect & 909.281 & 193.281 & 1049793.44 & 796387.19 \\
10 tie \#10 & 381.437 & 116.812 & 1039548.75 & 799456.25 \\
11 tie \#11 & 835.812 & 58.812 & 1048703.75 & 799236.25 \\
12 tie \#12 & 1468.062 & 162.812 & 1060876.25 & 795276.25 \\
13 tie \#13 & 2611.156 & 66.968 & 1083768.12 & 793733.12
\end{tabular}

\section*{Transformation Coefficients:}

1st order Transformation of cuxsS29au87.img
Imagine Tabular Report
\begin{tabular}{rrr} 
Row & \multicolumn{1}{c}{\(\mathrm{X}^{\prime}\)} & \multicolumn{1}{c}{\(Y^{\prime}\)} \\
\hdashline 1 & -44973.377384 & 47565.946814 \\
2 & 0.049446 & -0.007538 \\
3 & -0.007563 & -0.049550
\end{tabular}

Rectification Report:
RMS error report
cuxsS29au87.img
Reported Total RMS error: 0.455788
Imagine Tabular Report
\begin{tabular}{|c|c|c|c|c|c|}
\hline Row & Point ID & X Residual & Y Residual & RMS Error & Contribution \\
\hline 1 & 1 West Side Inter & -0.148 & 0.214 & 0.260 & 0.571 \\
\hline 2 & 2 Mid Intersect & -0.037 & 0.648 & 0.649 & 1.425 \\
\hline 3 & 3 N -Mid Intersect & 0.718 & -0.572 & 0.918 & 2.015 \\
\hline 4 & 4 W Side Main Rd & -0.136 & -0.241 & 0.276 & 0.606 \\
\hline 7 & 7 North of Fork & -0.051 & -0.359 & 0.363 & 0.796 \\
\hline 10 & tie \#10 & 0.034 & -0.011 & 0.036 & 0.079 \\
\hline 11 & tie \#11 & -0.003 & -0.119 & 0.119 & 0.261 \\
\hline 12 & tie \#12 & -0.428 & 0.345 & 0.550 & 1.206 \\
\hline 13 & tie \#13 & 0.050 & 0.095 & 0.107 & 0.235 \\
\hline Count & 9 & 9 & 9 & 9 & 9 \\
\hline Mean & N/A & 0.000 & -0.000 & 0.364 & 0.799 \\
\hline Minimum & N/A & -0.428 & -0.572 & 0.036 & 0.079 \\
\hline Maximum & N/A & 0.718 & 0.648 & 0.918 & 2.015 \\
\hline Stddev & N/A & 0.306 & 0.375 & 0.291 & 0.637 \\
\hline
\end{tabular}

SPOT PAN. PRE-PROCESSING DETAILS
\begin{tabular}{|c|c|c|c|c|c|}
\hline \multicolumn{6}{|l|}{} \\
\hline \multicolumn{6}{|l|}{} \\
\hline \multicolumn{6}{|l|}{Sub-scene coordinates from full-scene image:} \\
\hline \multicolumn{2}{|r|}{ULY} & 3000 & LRY & 6005 & \\
\hline \multicolumn{6}{|l|}{GCP Table:} \\
\hline 1 & GCP \#1 & 763.37 & 1547.62 & 1046640 & 812280 \\
\hline 2 & GCP \#2 & 1107.87 & 1606.12 & 1049940 & 811220 \\
\hline 3 & GCP \#3 & 1461.12 & 1647.62 & 1053330 & 810260 \\
\hline 4 & GCP \#4 & 1605.12 & 1813.12 & 1054575 & 808400 \\
\hline 5 & GCP \#5 & 1777.38 & 1762.38 & 1056270 & 808620 \\
\hline 6 & GCP \#6 & 2034.12 & 1854.62 & 1058730 & 807325 \\
\hline 7 & GCP \#7 & 2050.12 & 1669.62 & 1059150 & 809160 \\
\hline 8 & GCP \#8 & 2439.38 & 2020.62 & 1062500 & 805060 \\
\hline 9 & GCP \#9 & 2813.87 & 2199.62 & 1065880 & 802730 \\
\hline 10 & GCP \#10 & 2635.62 & 1544.12 & 1065140 & 809470 \\
\hline 11 & GCP \#11 & 3152.51 & 2543.64 & 1068670 & 798850 \\
\hline 12 & GCP \#12 & 2561.87 & 815.30 & 1065550 & 816850 \\
\hline 13 & GCP \#13 & 2836.91 & 2947.78 & 1064920 & 795340 \\
\hline 14 & GCP \#14 & 2367.13 & 2686.21 & 1060730 & 798600 \\
\hline 15 & GCP \#15 & 338.41 & 2885.71 & 1040356 & 799705 \\
\hline 16 & GCP \#16 & 1230.07 & 2781.14 & 1049329 & 799392 \\
\hline 17 & GCP \#17 & 2004.09 & 2826.22 & 1056910 & 797792 \\
\hline 18 & GCP \#18 & 3247.05 & 2745.20 & 1069330 & 796723 \\
\hline
\end{tabular}

Transformation Coefficients:
1st order Transformation of cupn29au87.img
Imagine Tabular Report
\begin{tabular}{|c|c|c|}
\hline Row & X' & \(\mathrm{Y}^{\prime}\) \\
\hline 1 & -90071.823432 & 97426.571011 \\
\hline 2 & 0.098816 & -0.015012 \\
\hline 3 & -0.015497 & -0.098690 \\
\hline
\end{tabular}

\section*{Rectification Report:}

RMS error report
cupnN29au87.img
Reported Total RMS error: 0.632958
Imagine Tabular Report
Row Point ID X Residual Y Residual RMS Error Contribution
\begin{tabular}{|c|c|c|c|c|c|}
\hline 2 & GCP \#2 & -0.525 & -0.778 & 0.938 & 1.482 \\
\hline 7 & GCP \#7 & -0.758 & 0.763 & 1.075 & 1.699 \\
\hline 12 & GCP \#12 & 0.741 & -0.002 & 0.741 & 1.171 \\
\hline 13 & GCP \#13 & -0.301 & -0.366 & 0.474 & 0.748 \\
\hline 14 & GCP \#14 & 0.211 & 0.160 & 0.265 & 0.418 \\
\hline 15 & GCP \#15 & 0.337 & -0.073 & 0.345 & 0.545 \\
\hline 16 & GCP \#16 & 0.200 & 0.680 & 0.709 & 1.120 \\
\hline 17 & GCP \#17 & 0.088 & -0.300 & 0.312 & 0.494 \\
\hline 18 & GCP \#18 & 0.006 & -0.084 & 0.084 & 0.133 \\
\hline Count & 9 & 9 & 9 & 9 & 9 \\
\hline Mean & N/A & 0.000 & 0.000 & 0.549 & 0.868 \\
\hline Minimum & N/A & -0.758 & -0.778 & 0.084 & 0.133 \\
\hline Maximum & N/A & 0.741 & 0.763 & 1.075 & 1.699 \\
\hline Stddev & N/A & 0.460 & 0.489 & 0.334 & 0.527 \\
\hline
\end{tabular}

Erdas Imagine File Information

File Name : crpnNS29au87.img Last Modified : Fri Aug 26 18:44:51 1994
Number of Layers : 1
Layer Information:
```

Name : :Layer_1
Width : }500
Height : 5000
Type : Continuous
Block Width : 64
Block Height : }6
Pixel Depth :Unsigned 8-bit
Compression Type : None
Statistics :
Last Modified : Fri Aug 26 18:44:51 1994
Maximum Value : 255.000000
Minimum Value : 0.000000
Mean : 43.798342
Median : 42.000000
Mode : 0.000000
Standard Deviation : 38.235842
Projection Information :
Projection Zone : 0
Spheroid Name : WGS 84
Georeferenced to : Transverse Mercator
Map Information :

```
Upper Left center X : 1035000.000000
Upper Left center Y : 825000.000000
Lower Right center X: 1085000.000000
Lower Right center Y: 775010.000000
Pixel X size : 10.000000
Pixel Y size : 10.000000

\section*{APPENDIX V}

\section*{The backup archive on exabyte tape contains:}
block 1
rwxrwxr-x101/11 392059787 Aug 4 16:56 1994 cauca.img
block 2
rw-rw-r--133/26 \(r w-r w-r--133 / 26\) rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 \(r w-r w-r--133 / 26\) \(r w-r w-r--133 / 26\) rw-rw-r--133/26
1915 Aug 12 13:50 1994 cuxsN29au87.cff
block 5
rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 \(r w-r w-r--133 / 26\) rw-rw-r--133/26
block 6
rw-rw-r--133/26
block 7
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
block 8
rwxrwxr-x133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26
rwxrwxr-x133/26 rw-rw-r--133/26 rw-rw-r--132/26 rw-rw-r--132/26 rw-rw-r--132/26 rw-rw-r--132/26 rw-rw-r--132/26 rw-rw-r--132/26
51639203 Aug 29 15:28 1994 crxs/crxsN29au87.img
128 Aug 29 \(08: 04\)
1994
crxs/crxsNS.fls
76965023 Aug 29 \(16: 10\) 1994 crxs/crxsNS29au87.img 0 crxs/crxsNS29au87.info

38976293 Aug 29 13:06 1994 crxs/crxsS29au87.img
323 Aug 29 09:26 1994 cutm07au89 crp
587 Aug 29 08:58 1994 cutm07au89.crp
5821 Aug 5 18:36 1994 cutm07au89.gcc
46468238 Aug 26 09:13 1994 cutm07au89.img
1464 Aug 29 09:13 1994 cutm07au89.rep

4001 Aug 26 18:21 1994 cupnN29au87.cff
cupnN29au87.crp
14023 Aug 26 18:08 1994 cupnN29au87.gcc
19335920 Aug 26 18:07 1994 cupnN29au87.img
1431 Aug \(2911: 51\)
1994
cupnN29au87. rep

51639203 Aug 29 15:28 1994 crxs/crxsN29au87.img
128 Aug 29 08:04 1994 crxs/crxsNS.fls
910 Aug 29 16:13 1994 crxs/crxsNS29au87.info

910 Aug 16 08:57 \(1994 \mathrm{crtm} / \mathrm{crtm07au89.info}\) 241142659 Aug 26 16:42 1994 crtm/crtm08au89.img
```

| 17872421 | Aug 26 | $18: 55$ | 1994 | crpn/crpnN29au87.img |
| ---: | :--- | :--- | :--- | :--- | :--- |
| 120 | Aug 26 | $18: 43$ | 1994 | crpn/crpnNS.fls |
| 34452766 | Aug 27 | $11: 12$ | 1994 | crpn/crpnNS29au87.img |
| 910 | Aug 27 | $11: 13$ | 1994 | crpn/crpnNS29au87.info |

14313403 Aug 27 10:59 $1994 \mathrm{crpn} / \mathrm{crpnS29}$ au87.img
pn/crpnN29au8

```

0 Aug 16 09:03 1994 cabuyal/crocabu/
48 Aug 16 09:03 1994 cabuyal/crocabu/tic 16 Aug 16 09:03 1994 cabuyal/crocabu/bnd 96 Aug 16 09:03 1994 cabuyal/crocabu/tol 355 Aug 24 14:04 1994 cabuyal/crocabu/log 132 Aug 16 09:03 1994 cabuyal/crocabu/lab 144 Aug 16 09:03 1994 cabuyal/crocabu/cnt 116 Aug 16 09:03 1994 cabuyal/crocabu/cnx 96 Aug 16 09:03 1994 cabuyal/crocabu/pat 5708 Aug 16 09:03 1994 cabuyal/crocabu/arc 116 Aug 16 09:03 1994 cabuyal/crocabu/arx 216 Aug 16 09:03 1994 cabuyal/crocabu/pal 116 Aug 16 09:03 1994 cabuyal/crocabu/pax 9691 Aug 16 09:12 1994 cabuyal/crocabu.evs

0 Aug 24 14:24 1994 cabuyal/info/
4180 Aug \(2414: 241994\) cabuyal/info/arcdr9 432 Aug 19 16:25 1994 cabuyal/info/arc000nit 80 Aug 19 16:25 1994 cabuyal/info/arc000dat 576 Aug 19 16:25 1994 cabuyal/info/arc001nit 80 Aug 19 16:25 1994 cabuyal/info/arc001dat 576 Aug 19 16:25 1994 cabuyal/info/arc002nit 80 Aug 19 16:25 1994 cabuyal/info/arc002dat
```

rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
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rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rwxrwxr-x132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rw-rw-r--132/26
rwxrwxr-x133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
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rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rwxrwxr-x133/26
rw-rw-r--133/26
rwxrwxr-x133/26
rw-rw-r--133/26
rwxrwxr-x133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
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rw-rw-r--133/26
rw-rw-r--133/26
r--r--r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rwxrwxr-x133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rwxrwxr-x133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26
rw-rw-r--133/26

```
rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26
rwxrwxr-x133/26 rw-rw-r--133/26 rw-rw-r--133/26
rwxrwxr-x133/26 rw-rw-r--133/26
rwxrwxr-x133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26
rwxrwxr-x133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26 rw-rw-r--133/26
\begin{tabular}{rllll}
3166404 & Aug & 29 & \(16: 51\) & 1994 \\
cabuyal/srtm/unsuper/rat-tass.img \\
3181145 & Aug & 24 & \(18: 47\) & 1994 \\
cabuyal/srtm/unsuper/hml.img \\
29766 & Aug & 25 & \(09: 34\) & 1994 \\
cabuyal/srtm/unsuper/rat-tass.sig \\
57293 & Aug & 12 & \(16: 16\) & 1994 \\
cabuyal/srtm/unsuper/rat-unclas.sig \\
3169187 & Aug & 19 & \(10: 08\) & 1994 \\
cabuyal/srtm/unsuper/rat-smth.img \\
3003501 & Aug & 17 & \(11: 50\) & 1994 \\
cabuyal/srtm/unsuper/rat.dat \\
1587833 & Aug & 19 & \(11: 11\) & 1994 \\
cabuyal/srtm/unsuper/rat-denx.img \\
1592655 & Aug & 19 & \(10: 59\) & 1994 \\
cabuyal/srtm/unsuper/rat-cplx.img
\end{tabular}
3166404 Aug 29 16:51 1994 cabuyal/srtm/unsuper/rat-tass.img
    3181145 Aug 24 18:47 1994 cabuyal/srtm/unsuper/hml.img
    29766 Aug 25 09:34 1994 cabuyal/srtm/unsuper/rat-tass.sig
    57293 Aug 12 16:16 1994 cabuyal/srtm/unsuper/rat-unclas.sig
3169187 Aug 19 10:08 1994 cabuyal/srtm/unsuper/rat-smth.img
1587833 Aug \(1911: 111994\) cabuyal/srtm/unsuper/rat-denx.img
1592655 Aug 19 10:59 1994 cabuyal/srtm/unsuper/rat-cplx.img
        - Aug 24 19:16 1994 cabuyal/srtm/unsuper/plots/
        402 Aug 24 19:16 1994 cabuyal/srtm/unsuper/plots/hml.map.ovr
            0 Aug 29 16:41 1994 cabuyal/srxs/
            0 Aug 24 14:24 1994 cabuyal/test/
            48 Aug 24 14:24 1994 cabuyal/test/tic
            16 Aug 24 14:24 1994 cabuyal/test/bnd
            5 Aug 24 14:24 1994 cabuyal/test/log
            cabuyal/vias/
            48 Aug 16 09.03 1994 cabuyal/vias/tic
            cabuyal/vias/bnd
    1444 Aug 24 14:30 1994 cabuyal/vias/log
        100 Aug 16 09:03 1994 cabuyal/vias/lab
        352 Aug 16 09:03 1994 cabuyal/vias/pat
            356 Aug 16 09:03 1994 cabuyal/vias/arc
            932 Aug 16 09:03 1994 cabuyal/vias/arx
            6240 Aug 16 09:03 1994 cabuyal/vias/aat
            696 Aug 16 09:03 1994 cabuyal/vias/nrf
    10705 Aug 24 15:59 1994 cabuyal/vias.evs```

