## Image Transforms

- Single band: spatial domain to frequency domain
- Multiple bands: spectral enhancement
- The enhancement techniques that require more than one band of data.
- Purposes:
- extract new bands of data that are more interpretable to the eye (VI, Tasseled Cap)
- apply mathematical transforms and algorithms (Band ratioing)
- display a wider variety of information in the three available color guns (R, G, B) (PCA)
- compress bands of data that are similar (PCA)


## Indices

- Band ratioing
- NDVI (Normalized Difference Vegetation Index)
$\frac{I R-R}{I R+R}$



## Spectral Properties of Objects



Figure 1.10 Typical spectral reflectance curves for vegetation, soil, and water.

## Soil Reflectance Separation




Mixture of Soil \& Vegetation Reflectance






## Dehazing Algorithm

- Tasseled Cap Transformation
- Subtracts Haze from the blue band

```
Brightness = .3037(TM1) +.2793)(TM2) +.4743 (TM3) +.5585(TM4) +. 5082 (TM5)
    +.1863 (TM7)
Greenness = -.2848(TM1) -.2435(TM2) -.5436(TM3) +.7243(TM4) +.0840 (TM5)
    -. 1800 (TM7)
Wetness = .1509(TM1)+.1973 (TM2) +. .3279(TM3) +.3406 (TM4)-.7112 (TM5)
    - .4572 (TM7)
Haze = .8832(TM1) -.0819 (TM2) -. 4580 (TM3) -.0032 (TM4) -.0563 (TM5) +
    .0130 (TM7)
```



## RGB / IHS Transform



IHS Color Coordinate System


## Data Fusion

- Fuse data
- Resolution merge
- Merge SPOT 10m pan with 20m multi-spectral
- Sensor merge
- Merge Radar intensity with multi-spectral


## Fusion Procedures

1. Band transformation

- RGB - IHS
- PCA, Tasseled Cap
- Wavelet Analysis

2. Band replacement
3. Band back-transformation


## TM multi-spectral merged with Radar image



## Principal Components Analysis (aka Factor Analysis)

- Capture the main factors \& reduce data redundancy
- Based on covariance matrix
- Covariance measures the tendencies of data file values in the same pixel, but in different bands, to vary with each other, in relation to the means of their respective bands.

$$
\operatorname{var}(X)=\frac{\sum_{i=1}^{n}\left(X_{i}-\bar{X}\right)\left(X_{i}-\bar{X}\right)}{(n-1)} \quad \operatorname{cov}(X, Y)=\frac{\sum_{i=1}^{n}\left(X_{i}-\bar{X}\right)\left(Y_{i}-\bar{Y}\right)}{(n-1)}
$$

[^0]
# Principal Components Analysis (PCA) 

Figure 6-18: Two Band Scatterplot



## Covariance Matrix

TABLE 10.1. Similarity Matrices for Seven Bands of a TM Scene

| Covariance matrix |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| 1. 48.8 | 29.2 | 43.2 | 50.0 | 76.5 | 0.9 | 44.9 |
| 2. 29.2 | 20.3 | 29.0 | 48.6 | 65.4 | 1.5 | 32.8 |
| 3. 43.2 | 29.0 | 46.4 | 59.9 | 101.2 | 0.6 | 53.5 |
| 4. 49.9 | 48.6 | 59.9 | 327.8 | 325.6 | 12.4 | 104.32 |
| 5. 76.5 | 65.4 | 101.2 | 325.6 | 480.5 | 10.2 | 188.5 |
| 6. 0.9 | 1.5 | 0.6 | 12.5 | 10.2 | 14.0 | 1.1 |
| 7. 45.0 | 32.8 | 53.5 | 104.3 | 188.5 | 1.1 | 90.8 |
| Correlation matrix |  |  |  |  |  |  |
| 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| 1. 1.00 |  |  |  |  |  |  |
| 2. 0.92 | 1.00 |  |  |  |  |  |
| 3. 0.90 | 0.94 | 1.00 |  |  |  |  |
| 4. 0.39 | 0.59 | 0.48 | 1.00 |  |  |  |
| 5. 0.49 | 0.66 | 0.67 | 0.82 | 1.00 |  |  |
| 6. 0.03 | 0.08 | 0.02 | 0.18 | 0.12 | 1.00 |  |
| 7. 0.67 | 0.76 | 0.82 | 0.60 | 0.90 | 0.02 | 1.00 |



## Calculating Components

- Component A

$$
A=C_{1} X_{1}+C_{2} X_{2}+C_{3} X_{3}+C_{4} X_{4}
$$

- Component coefficients
- $\mathrm{C}_{1}=0.35, \mathrm{C}_{2}=-0.08, \mathrm{C}_{3}=0.36, \mathrm{C}_{4}=0.86$
- BV of input bands
- $X_{1}=28, X_{2}=29, X_{3}=21, X_{4}=54$
- What is the BV of component $A$ ?


## Fourier Transformation



## Image Domains

- Spatial domain (2D images)
- Spectral domain (Feature space plots, scatterplots)
- Frequency domain (Fourier Transformation)



## Fourier Filtering

Brightness Image Fourier Transform Inverse Transformed


Low-Pass Filtered Inverse Transformed


High-Pass Filtered Inverse Transformed


Fourier Analysis (http://cns-alumni.bu.edu/~slehar/fourier/fourier.html)



Figure 63: Wavelet Resolution Merge



[^0]:    http://csnet.otago.ac.nz/cosc453/student_tutorials/principal_components.pdf

