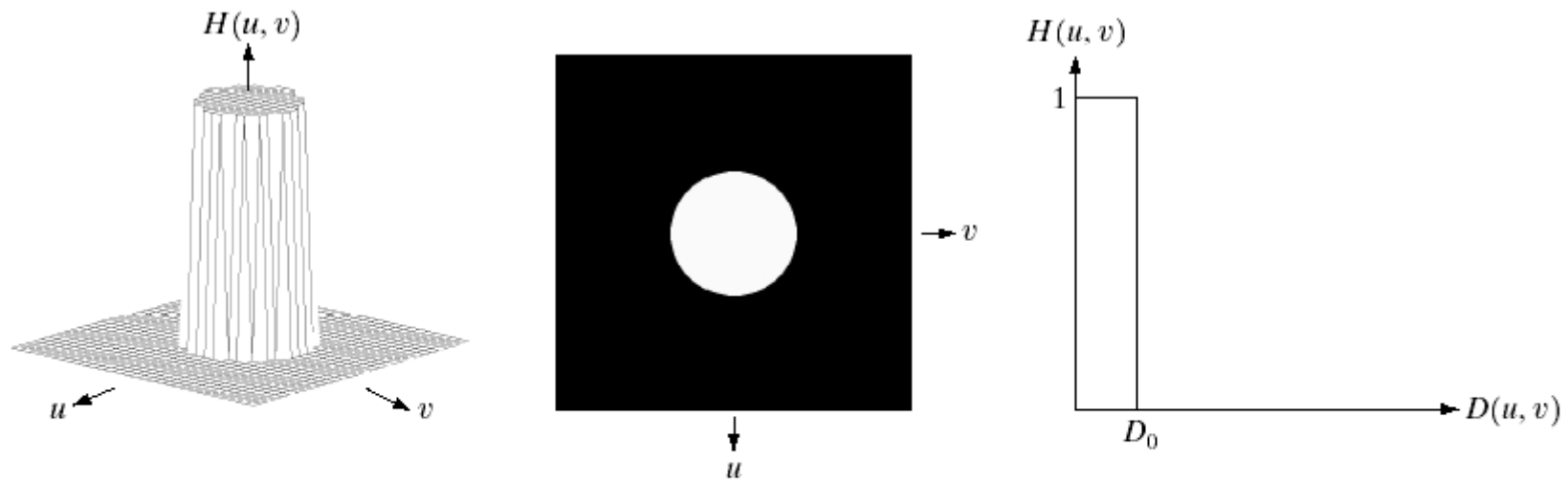


# Ideal lowpass filters

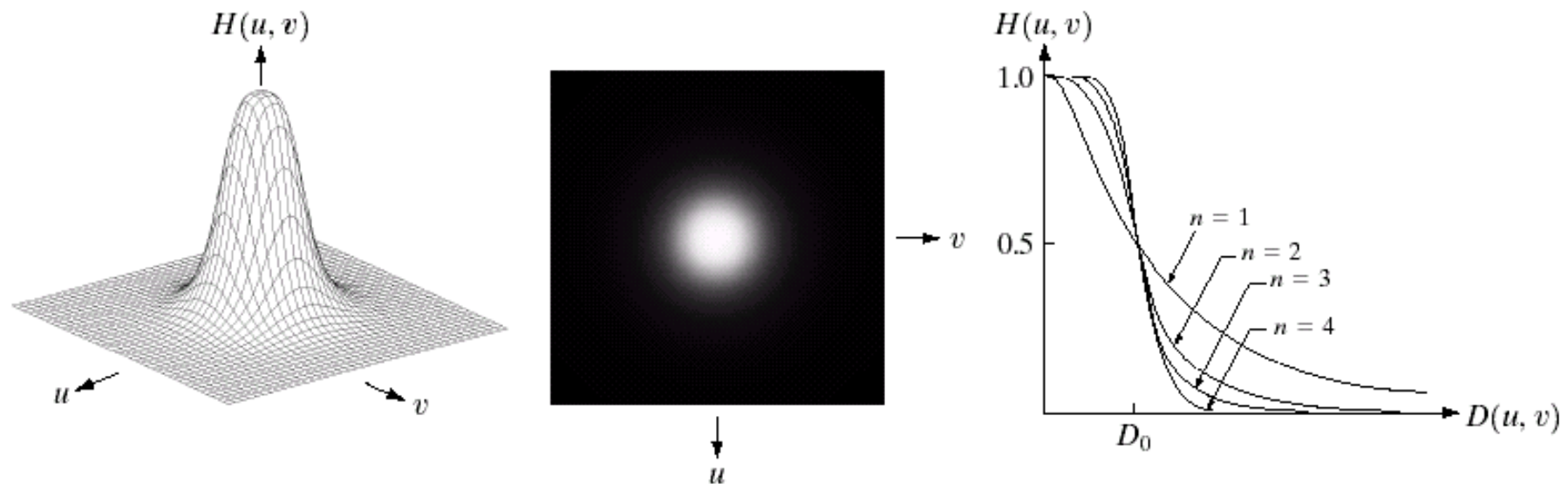


a b c

**FIGURE 4.10** (a) Perspective plot of an ideal lowpass filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross section.

---

# Butterworth Lowpass Filter



a b c

**FIGURE 4.14** (a) Perspective plot of a Butterworth lowpass filter transfer function. (b) Filter displayed as an image. (c) Filter radial cross sections of orders 1 through 4.

a
b c
d e

**FIGURE 3.44**  
 A  $3 \times 3$  region of an image (the  $z$ 's are gray-level values) and masks used to compute the gradient at point labeled  $z_5$ . All masks coefficients sum to zero, as expected of a derivative operator.

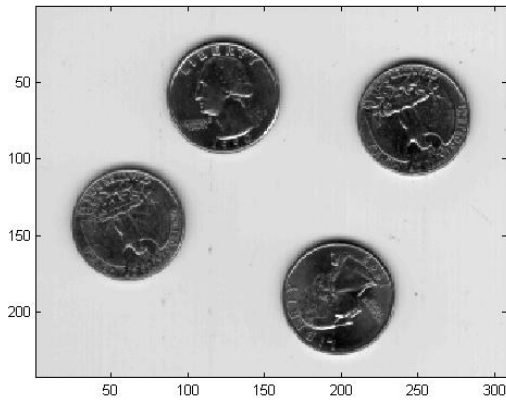
$z_1$	$z_2$	$z_3$
$z_4$	$z_5$	$z_6$
$z_7$	$z_8$	$z_9$

### Roberts Cross-Gradient

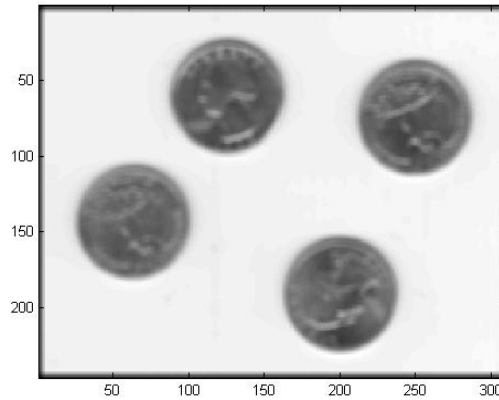
-1	0	0	-1
0	1	1	0

### Sobel Operators

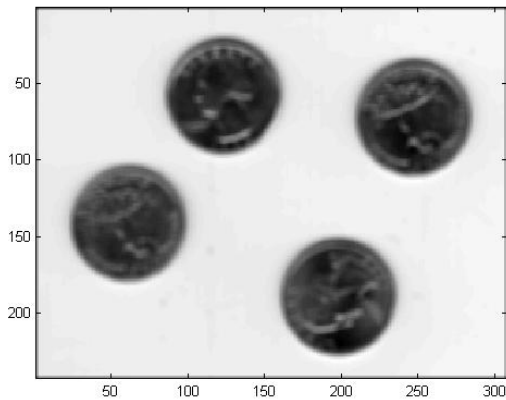
-1	-2	-1	-1	0	1
0	0	0	-2	0	2
1	2	1	-1	0	1



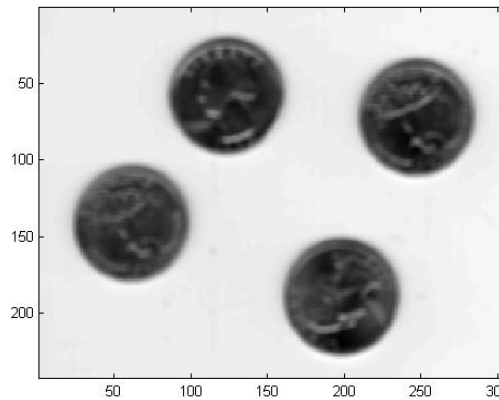
A (Original image)



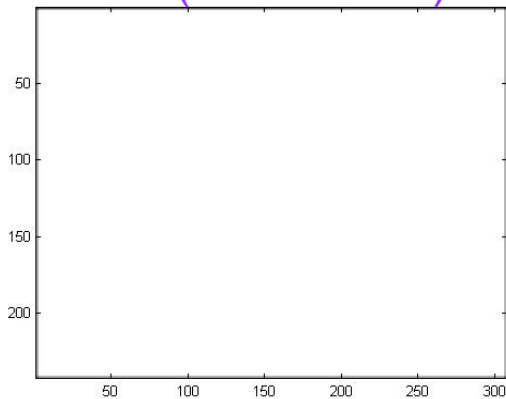
B (convolution)



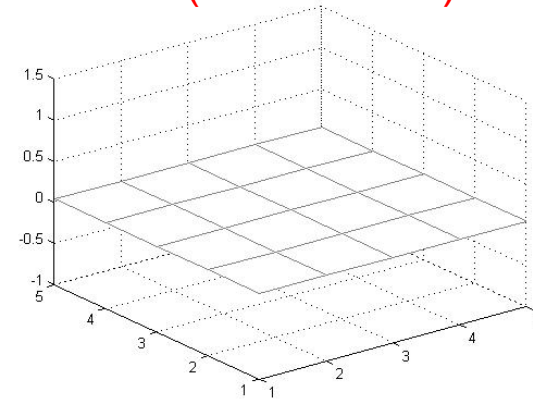
B (convolution)



B (convolution)



C1

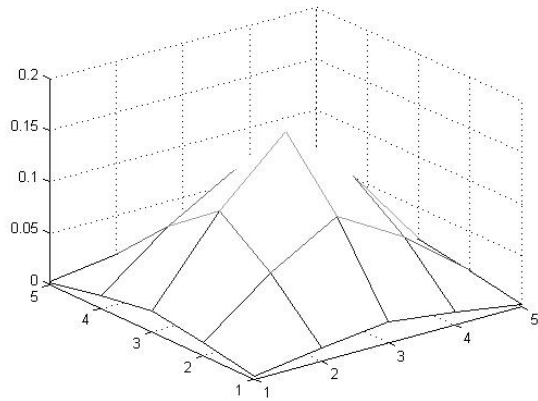


M

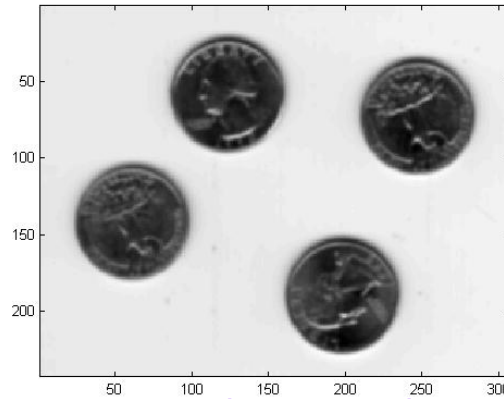
```

A=imread('eight.tif');
A=double(A);
imagesc(A)
colormap(gray)
M=ones(5)/25;
B=conv2(A,M);
figure
imagesc(B)
colormap(gray)
B=conv2(A,M,'same');
imagesc(B)
colormap(gray)
C=ones(size(B));
B=conv2(A,M,'same')./
conv2(C,M,'same');
imagesc(B)
colormap(gray)
C1=conv2(C,M,'same');
figure
colormap(gray)
imagesc(C1*255)
mesh(M)

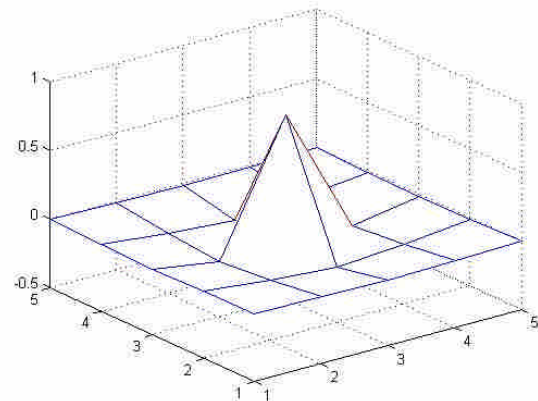
```



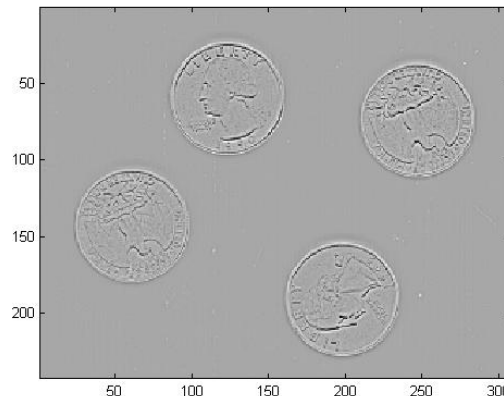
M1



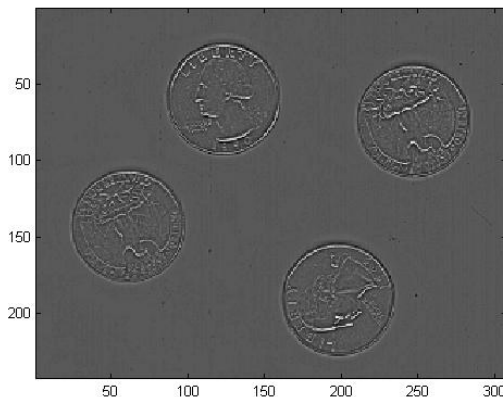
B (lowpass)



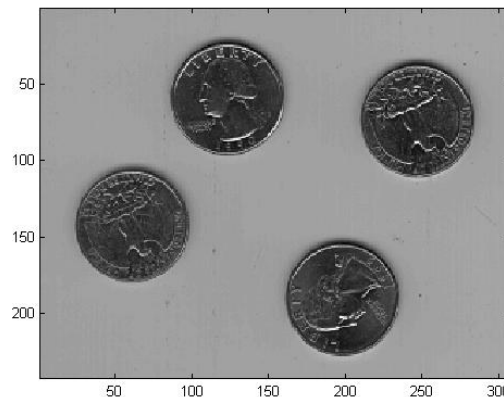
M2



B (highpass)



C (highpass=1-lowpass)



C+A (highboost)

```
f=[.05 .25 .4 .25 .05];
M1=f'*f;
mesh(M1)
B=conv2(A,M1,'same')./
conv2(C,M1,'same');
figure
imagesc(B)
colormap(gray)
D=zeros(5);
D(3,3)=1;
M2=D-M1;
mesh(M2)
B=conv2(A,M2,'same')./
conv2(C,M2,'same');
imagesc(B)
B=conv2(A,M1,'same')./
conv2(C,M1,'same');
C=A-B;
imagesc(C)
imagesc(C+A)
```

# Butterworth Lowpass Filter

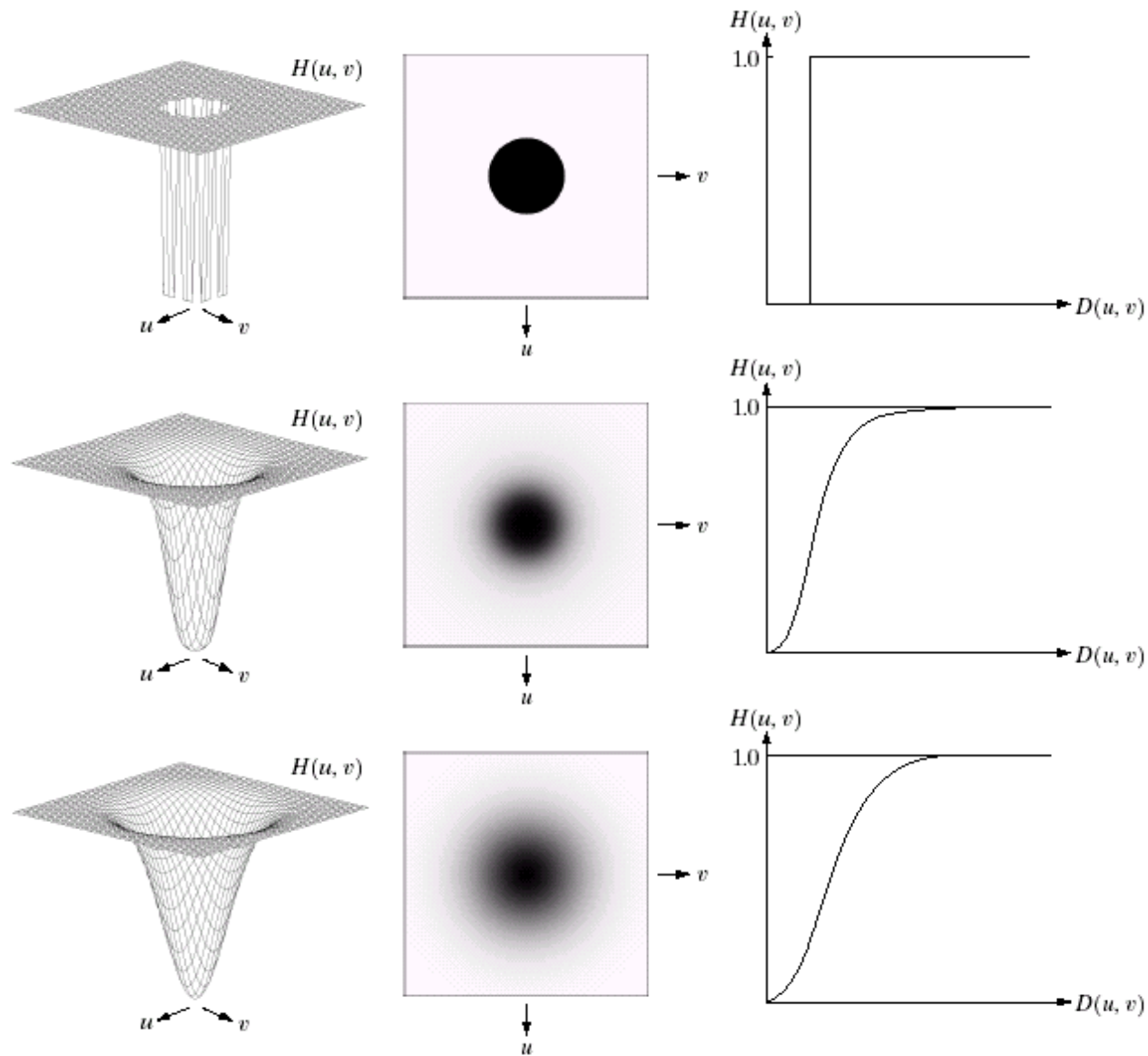
- This filter does not have a sharp discontinuity establishing a clear cutoff between passed and filtered frequencies.

$$H(u, v) = \frac{1}{1 + [D(u, v) / D_0]^{2n}}$$

where  $D_0$  is the cutoff frequency , and  $D(u, v)$  is

$$D(u, v) = [(u - M / 2)^2 + (v - N / 2)^2]^{1/2}$$

- This is more appropriate for image smoothing than the ideal LPF, since this does not introduce ringing.



a	b	c
d	e	f
g	h	i

**FIGURE 4.22** Top row: Perspective plot, image representation, and cross section of a typical ideal highpass filter. Middle and bottom rows: The same sequence for typical Butterworth and Gaussian highpass filters.

# Gaussian Lowpass Filter

- Gaussian lowpass filter is defined by

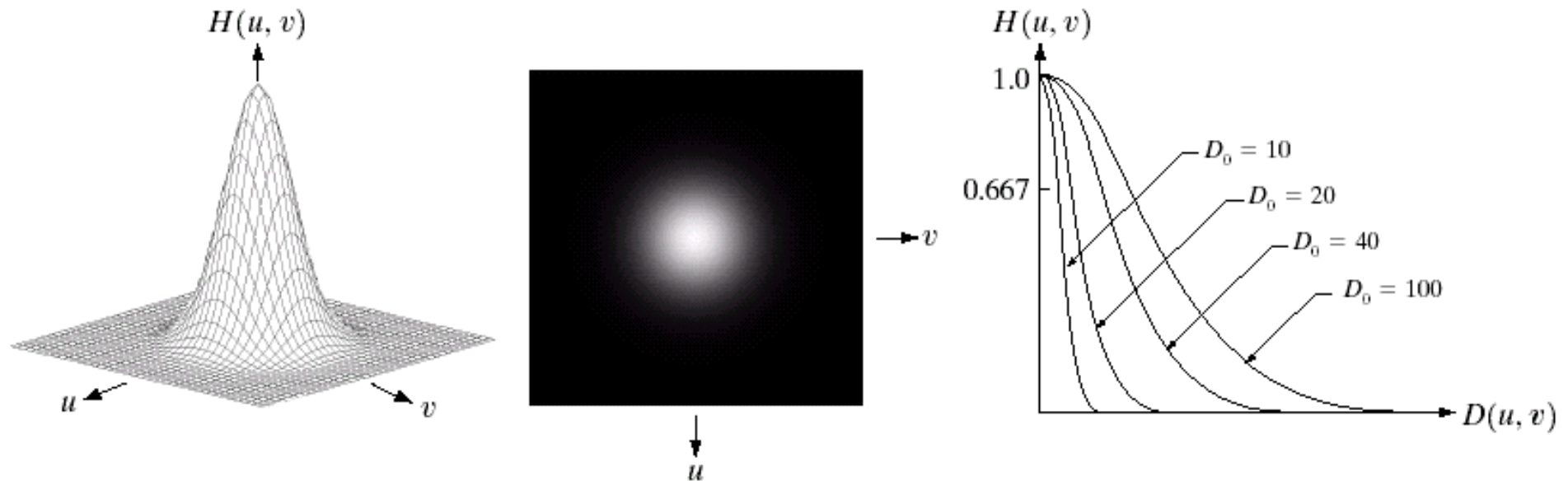
$$H(u, v) = e^{-D^2(u, v) / 2\sigma^2}$$

$D(u, v)$  is the distance from the origin of the Fourier transform.  
by letting  $\sigma = D_0$ , we have

$$H(u, v) = e^{-D^2(u, v) / 2D_0^2}$$

where  $D_0$  is the cutoff frequency.

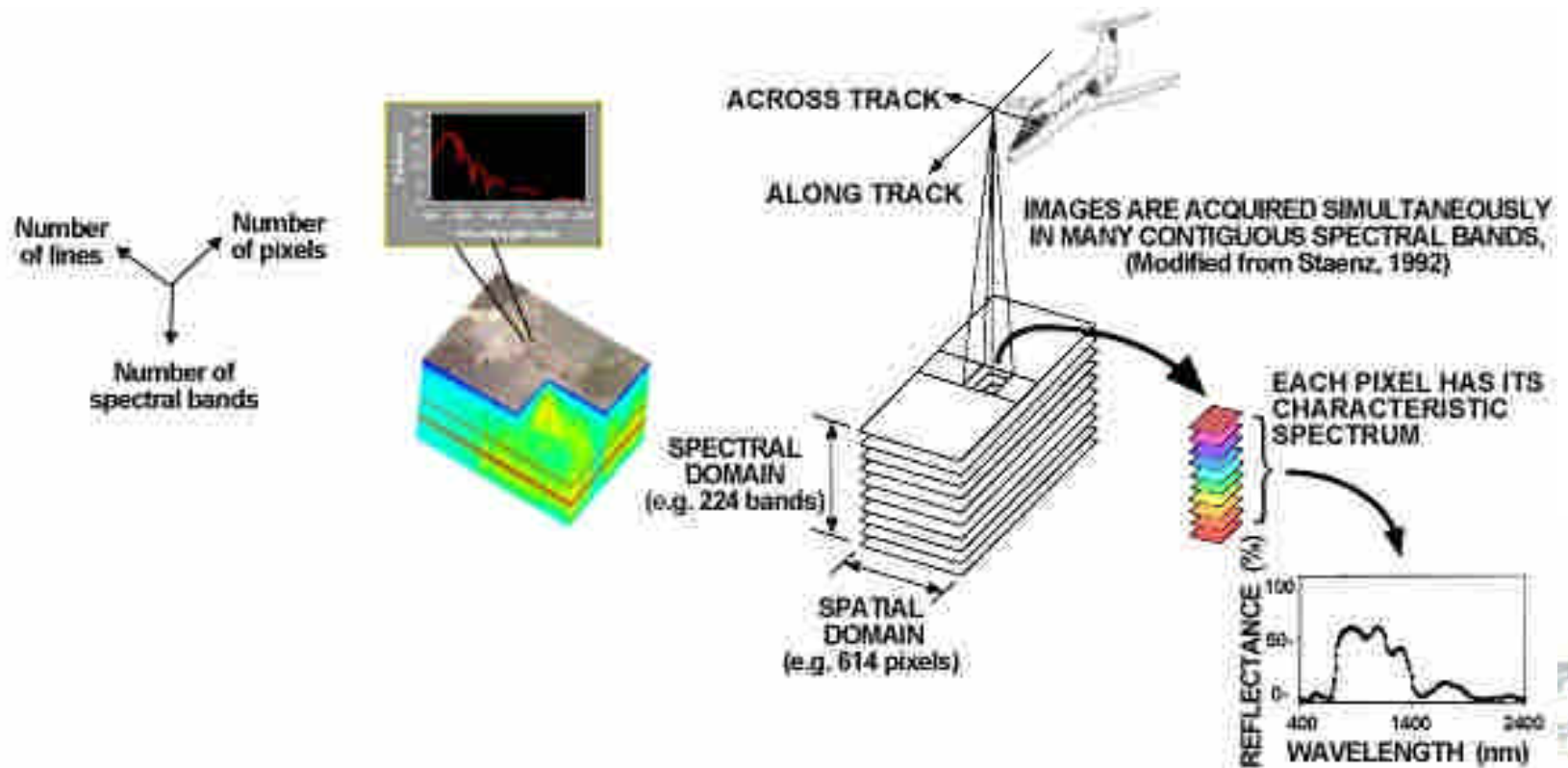
# Gaussian Lowpass Filter



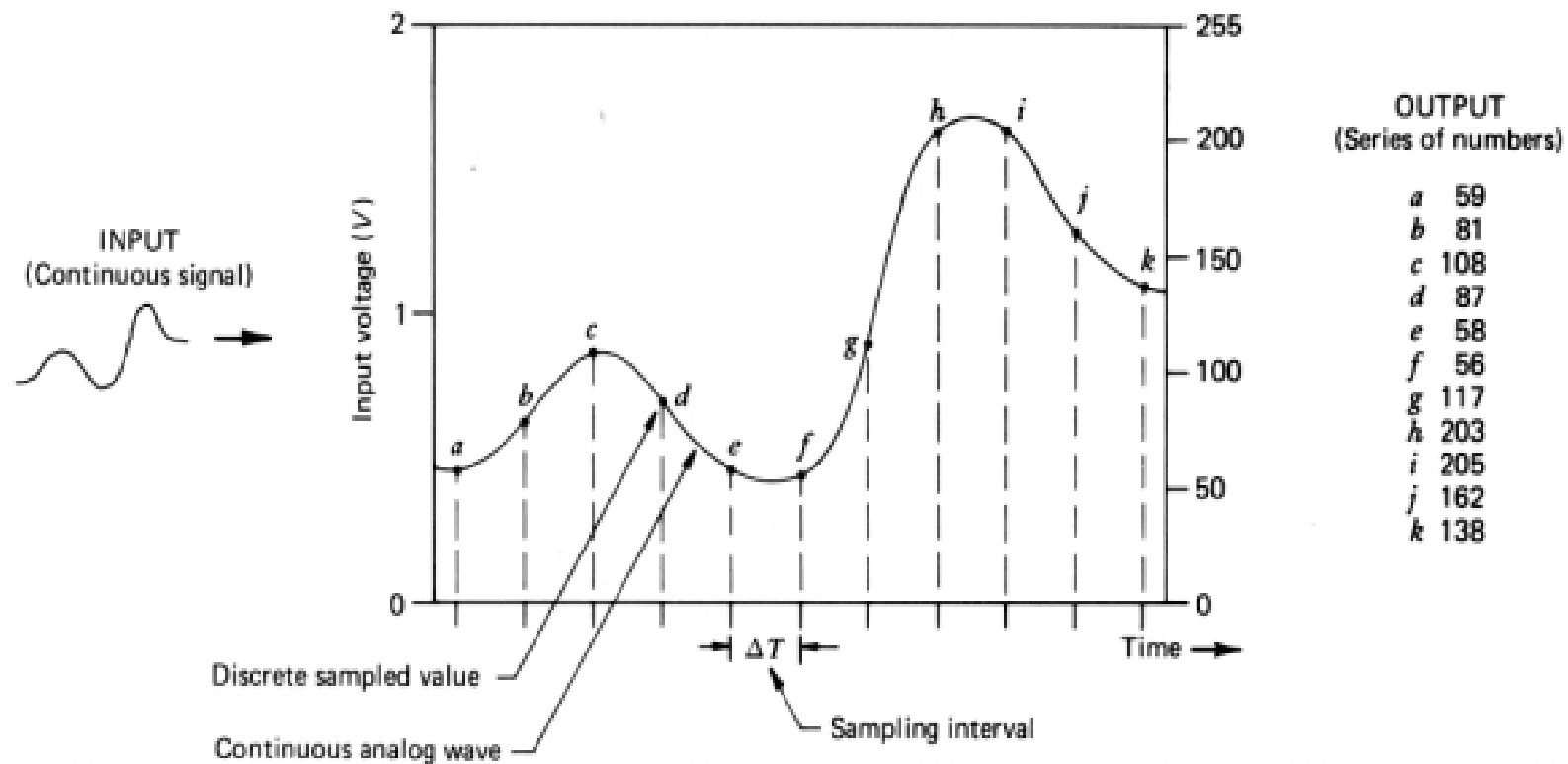
a b c

**FIGURE 4.17** (a) Perspective plot of a GLPF transfer function. (b) Filter displayed as an image. (c) Filter radial cross sections for various values of  $D_0$ .

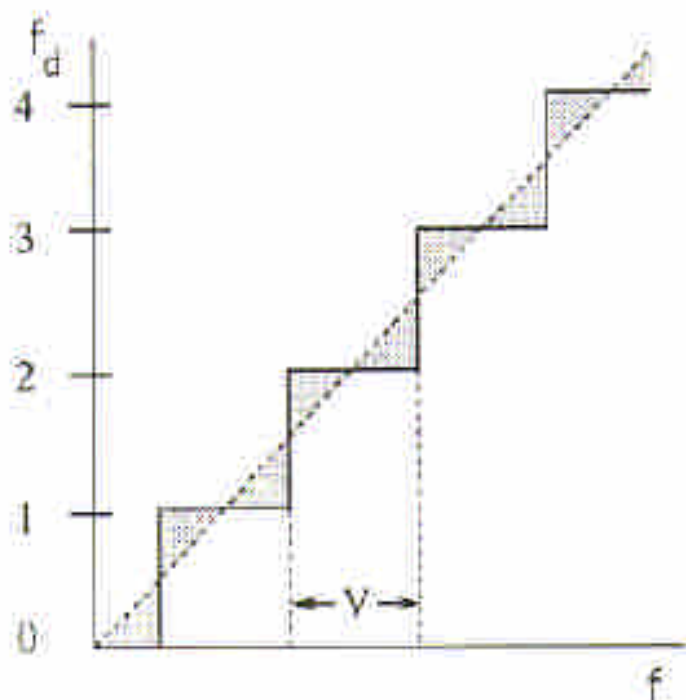
# Digital Remote Sensing Image



# Radiometric Resolution



# Quantization



$f$  : analogue intensity

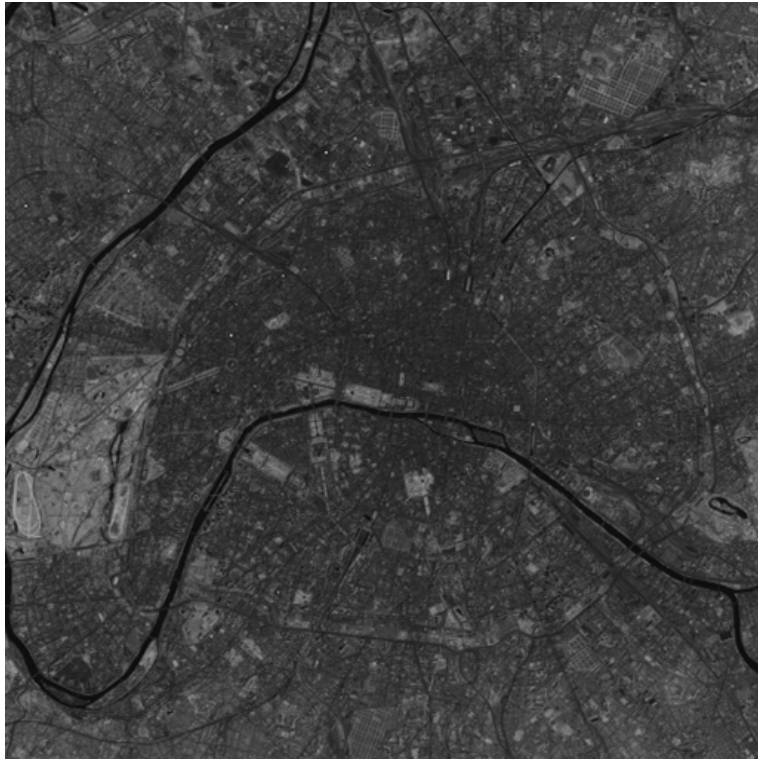
$f_d$  : quantized intensity

$V$  : unit intensity

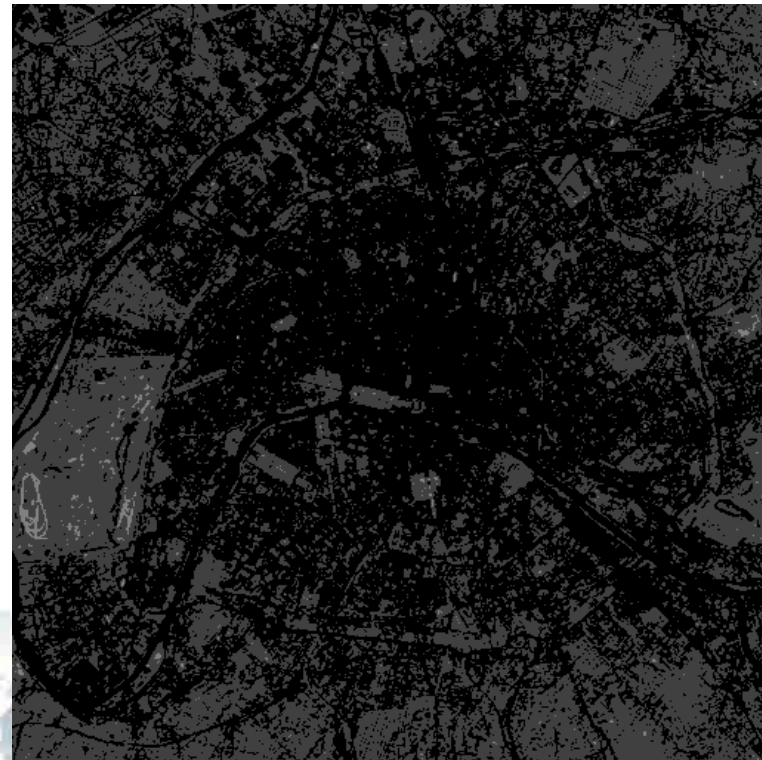
$n$  : integer

$(n-0.5)V \leq f < (n+0.5)V \rightarrow f_d = n$

quatization error =  $f - f_d$  (shaded part)

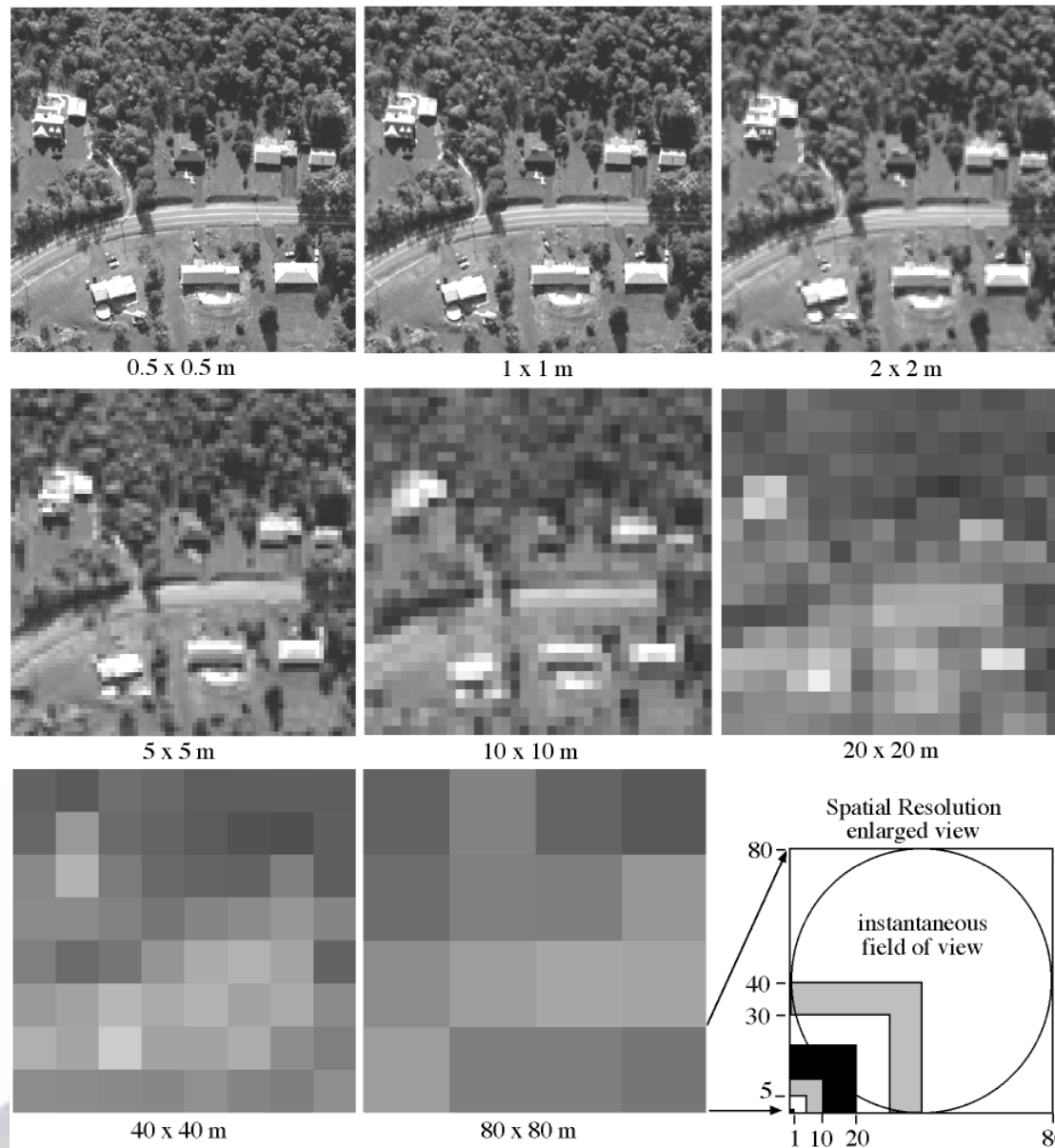


8 bits

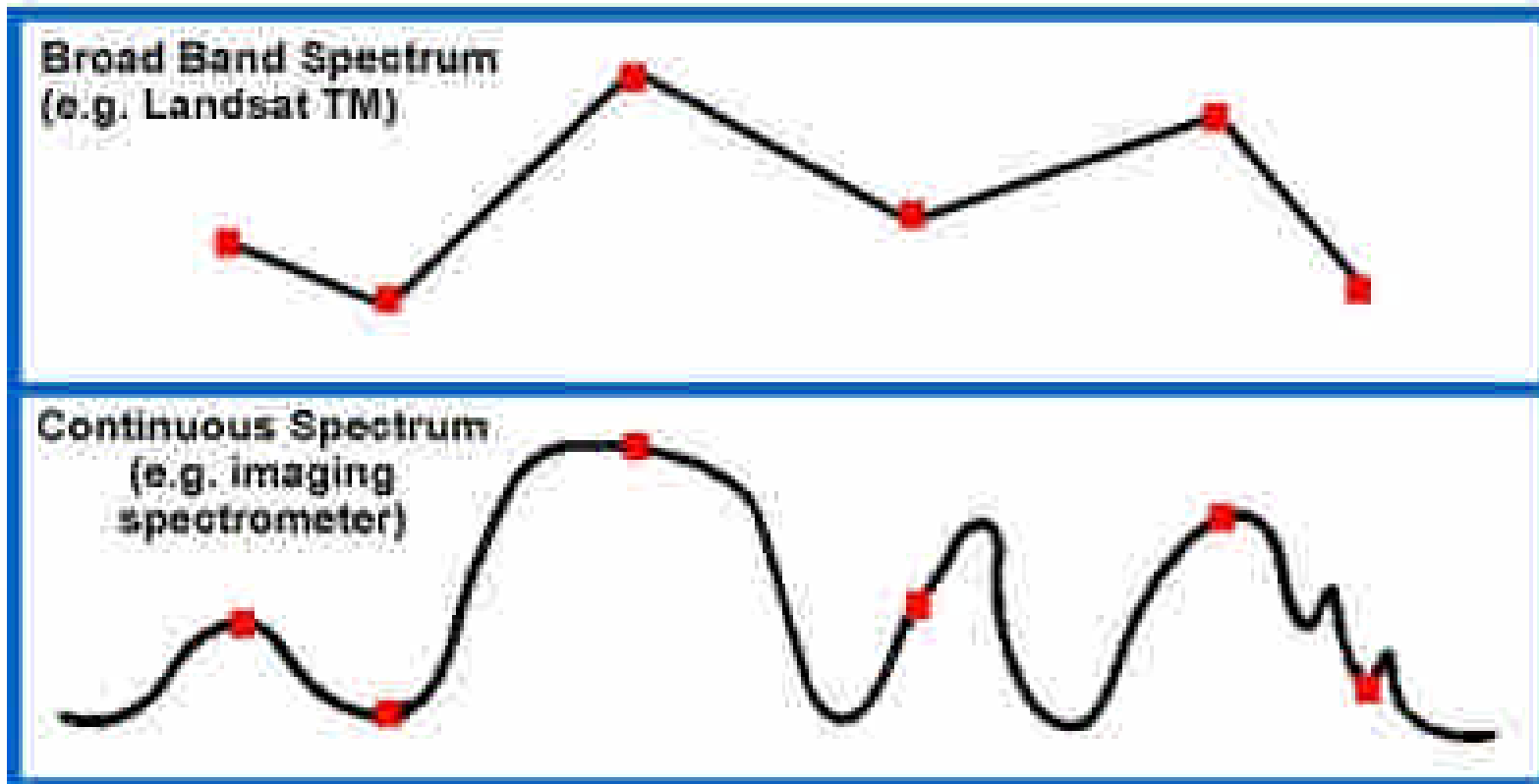


2 bits

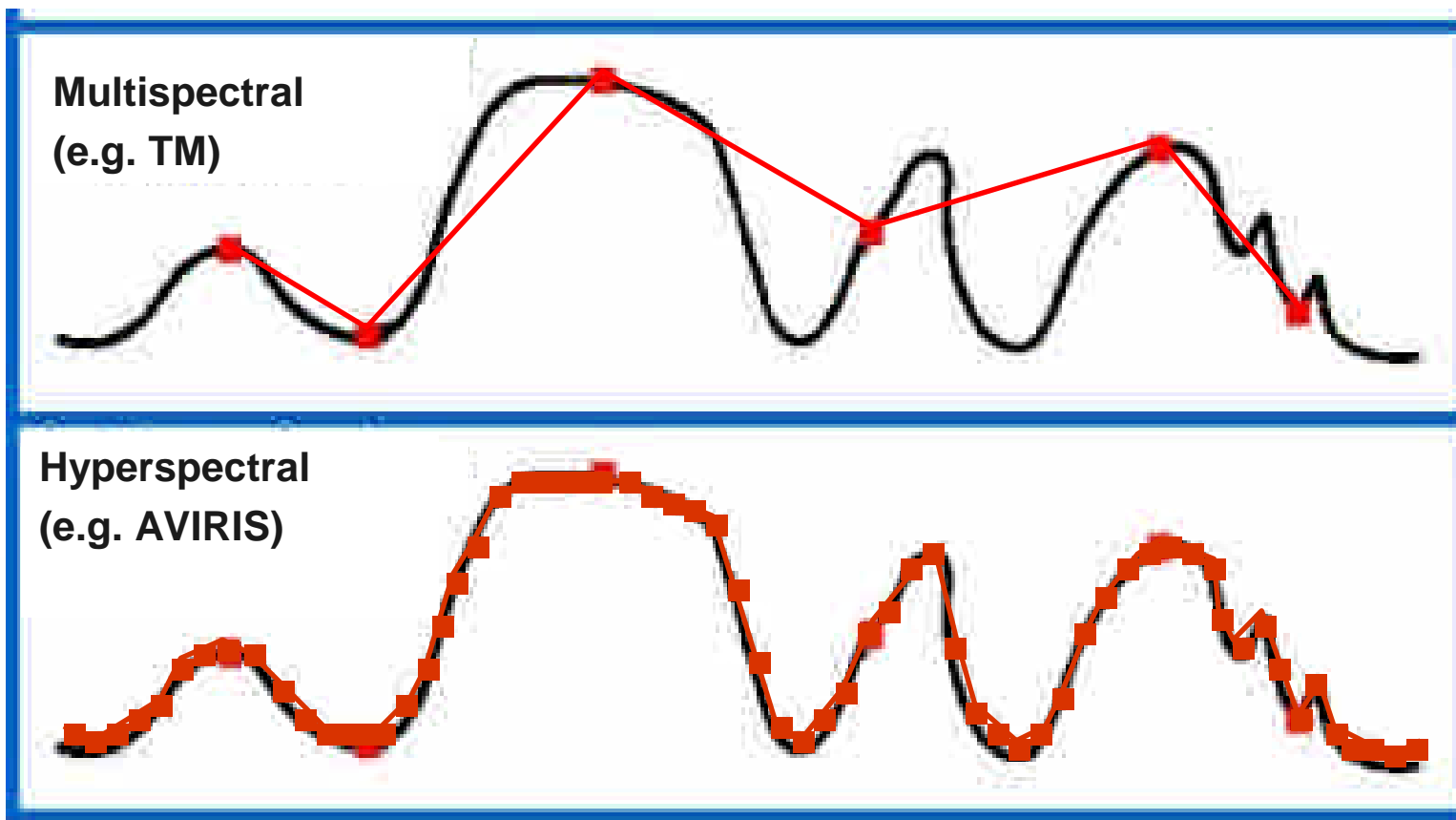
# Spatial Resolution

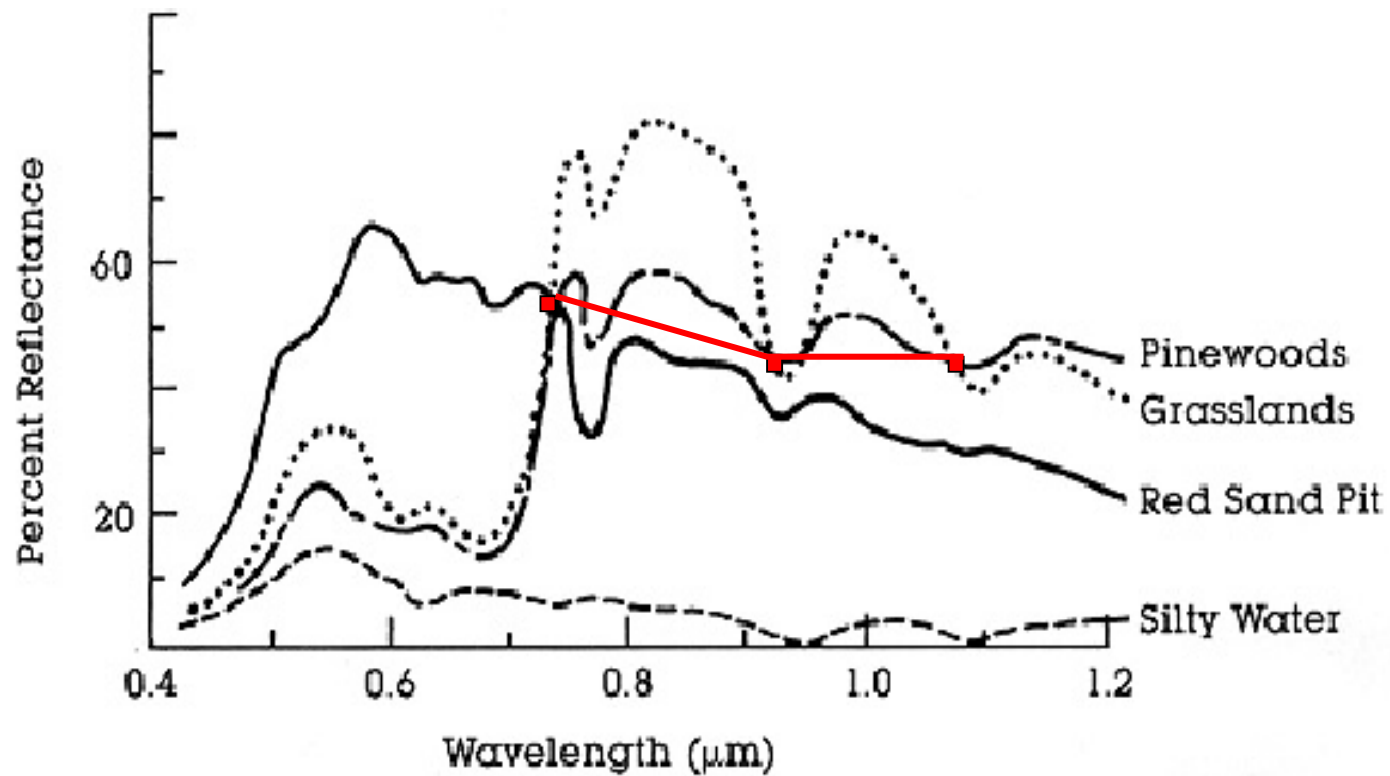


# Spectral Resolution

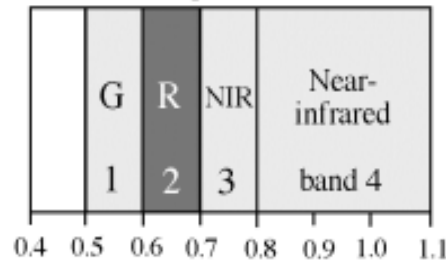


# Multispectral vs Hyperspectral

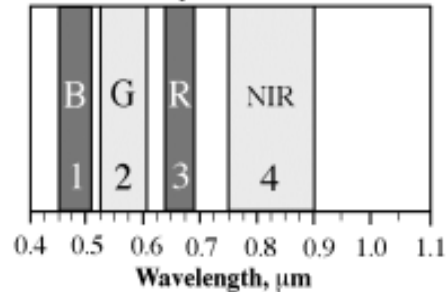




**Landsat Multispectral Scanner (MSS)**



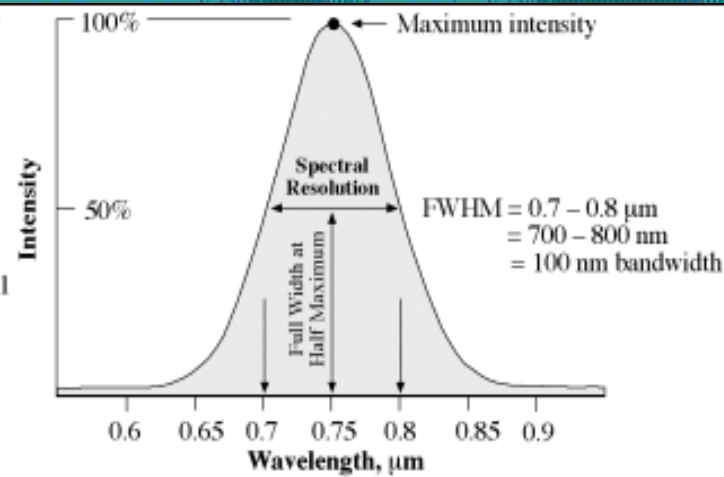
**Positive Systems ADAR 5500**



a. Nominal spectral resolution of the Landsat Multispectral Scanner and Positive Systems ADAR 5500 digital frame camera.



c. Single band of ADAR 5500 data



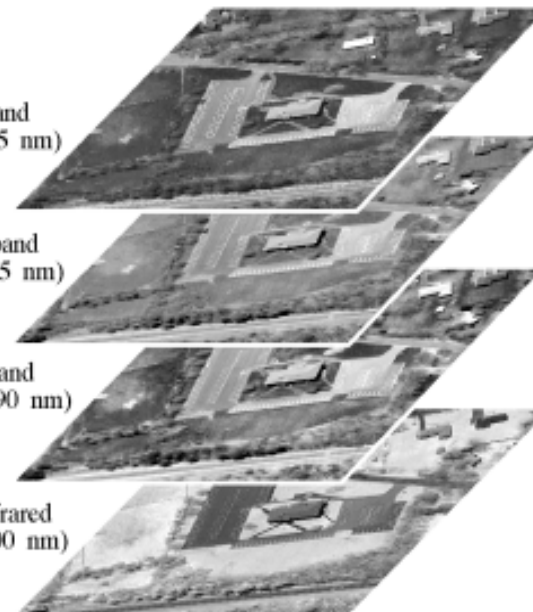
b. Precise bandpass measurement of a detector based on Full Width at Half Maximum (FWHM) criteria

blue band  
(450 – 515 nm)

green band  
(525 – 605 nm)

red band  
(640 – 690 nm)

near-infrared  
(750 – 900 nm)



d. Multispectral remote sensing



## Temporal Resolution

System	Resolution (day)
Landsat 1, 2, 3	18
Landsat 4, 5	16
AVHRR	1
SPOT	26
MODIS	2

# Data Compression

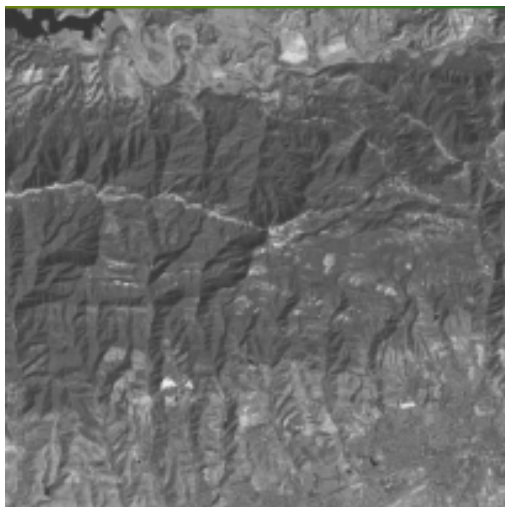
## ■ SPOT5

- XI: XS1, XS2, XS3: 10m
- SWIR: 20m
- PAN: 5m
- 60km x 60km:
  - XI:  $(60000(\text{m})/10(\text{m}))^2 \times 3(\text{band}) = 108 \text{ MB}$
  - SWIR:  $(60000(\text{m})/20(\text{m}))^2 = 9 \text{ MB}$
  - PAN:  $(60000(\text{m})/5(\text{m}))^2 = 144 \text{ MB}$
  - Total: 261 MB

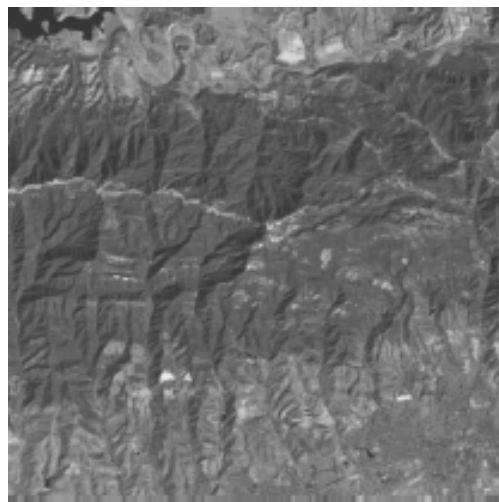
<b>COMPRESSION TYPE</b>
<i>Lossless</i>
<b>Run Length Coding</b>
<b>Huffman Coding</b>
<i>Lossy</i>
<b>Joint Photographic Experts Commission (JPEG) - DCT</b>
<b>Vector Quantization (VQ)</b>



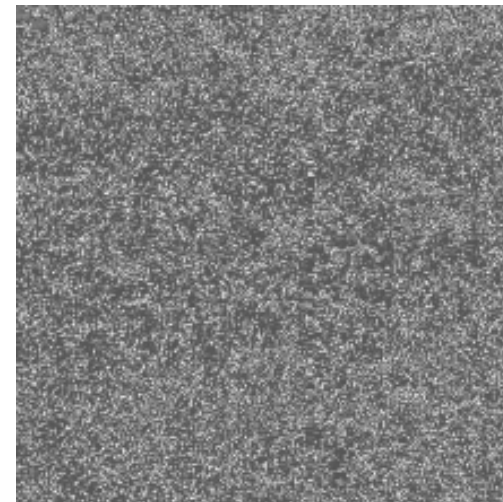
# Vector Quantization



Original



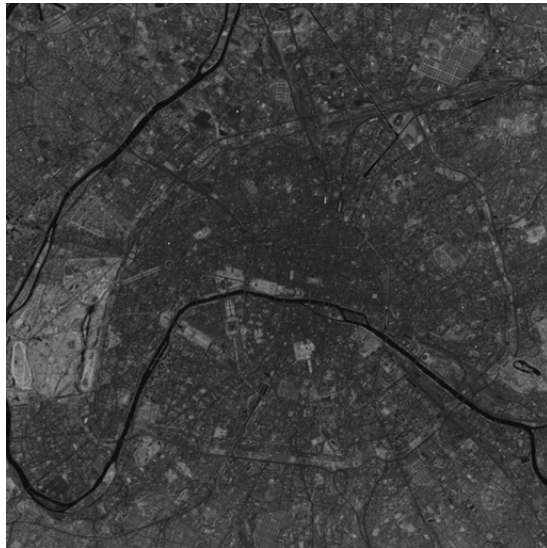
10:1 ratio



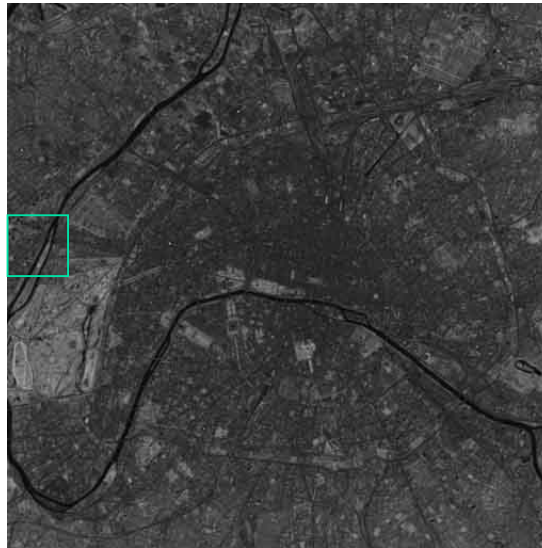
Error



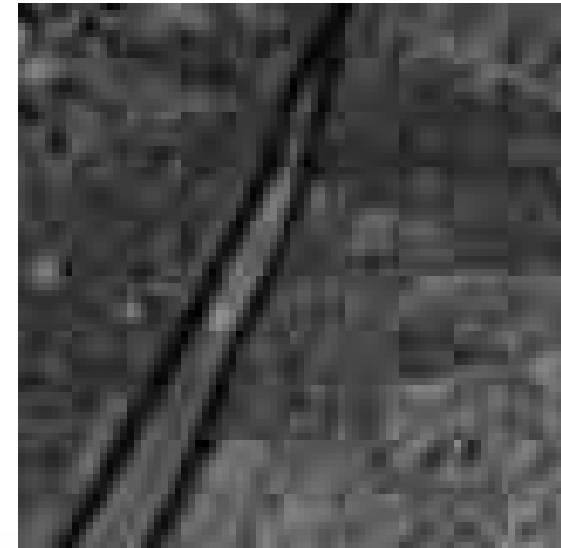
# JPEG



Original



8:1 ratio

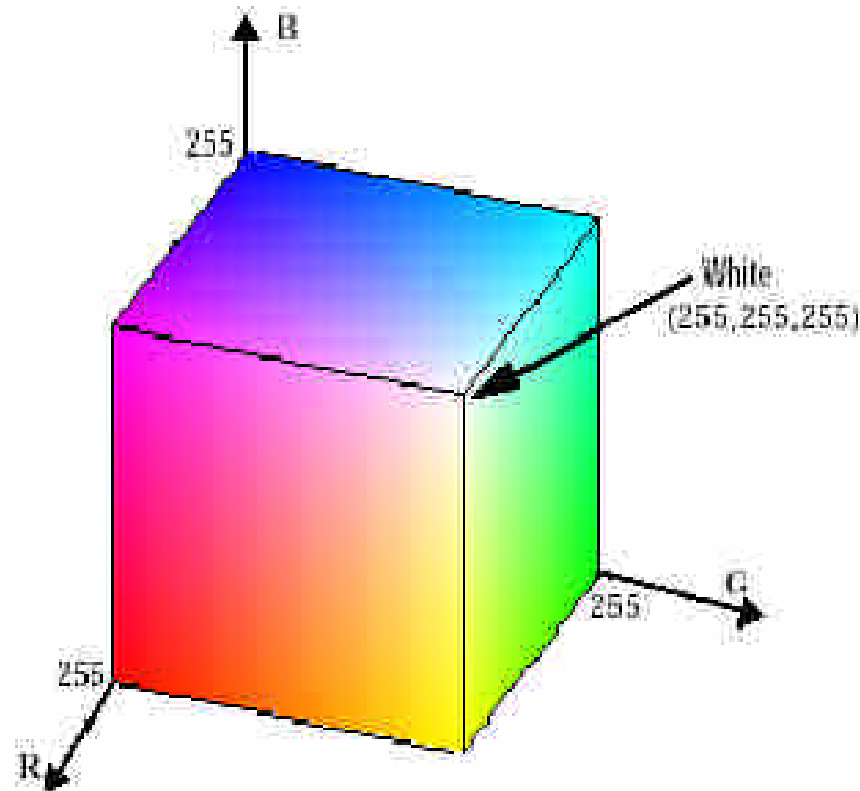


## Image Display

- False color
- Pseudo color
- Color system

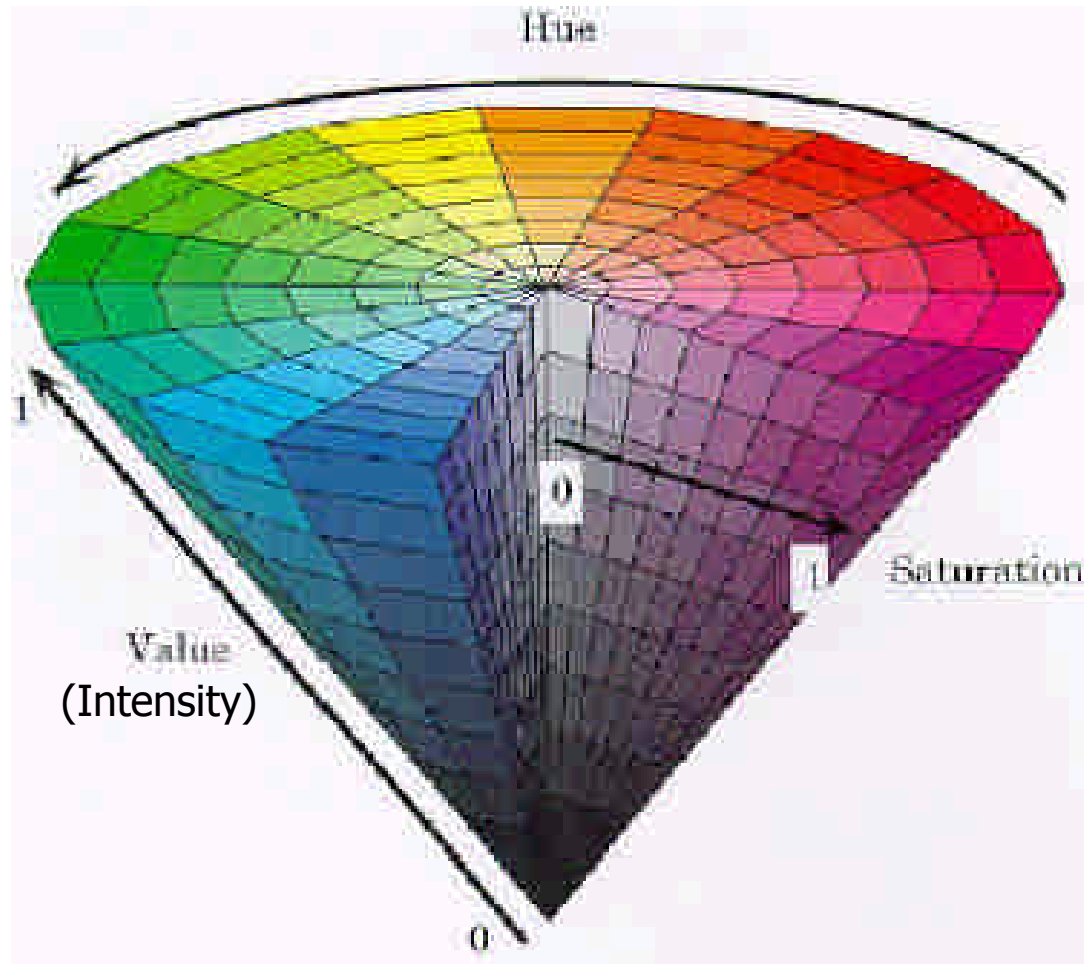


# Color



**Color coordinate**

# Hue, Saturation, Value (HSV) (色調, 飽和度, 亮度)



# 亮度變化

Change in Intensity

original



40% intensity



160% intensity

# 飽和度變化

Change in Saturation



original



40% saturation



160% saturation

# 色調變化

Change in Hue



original



-60 degree



-120 degree



-180 degree

# 其他色彩系統

- YIQ:
  - 電視系統
- YUV:
  - PAL 和數位影片
- CMY:
  - 印表機系統

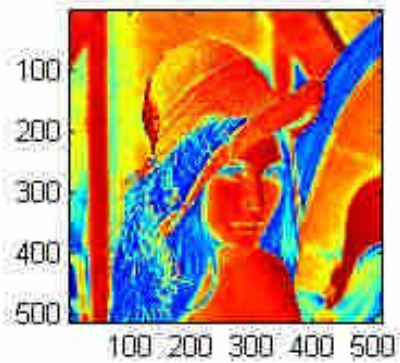
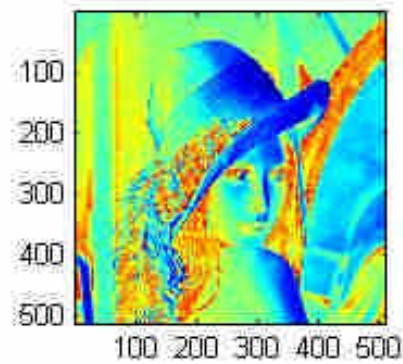
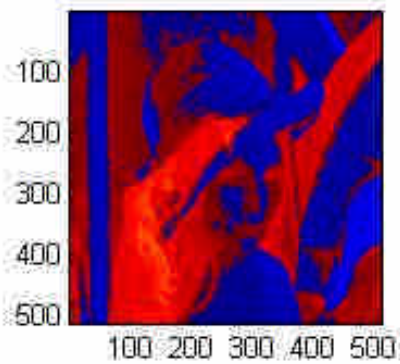
# 色彩系統轉換

- **RGB to CMY**
  - $C = W - R$
  - $M = W - G$
  - $Y = W - B$
- **RGB to YIQ**
  - $Y = 0.299R + 0.587G + 0.114B$
  - $I = 0.596R - 0.274G - 0.322B$
  - $Q = 0.211R - 0.523G + 0.312B$
- **RGB to HSI**
  - $H = (90 - \arctan(F/1.71) + C) / 360$ , with  $F = (2R - G - B)/(G - B)$   
and  $C = 0$  if  $G > B$ , or  $C = 180$ , if  $G < B$
  - $S = 1 - \min(R, G, B) / V$ ,
  - $I = (R + G + B) / 3$ .
- **RGB to YUV**
  - $Y = 0.299R + 0.587G + 0.114B$ , Intensity, Luminance
  - $U = 0.493 ( B - Y )$ ,
  - $V = 0.877 ( R - Y )$ , Color Differences, Chrominance



lena

```
A=imread('lena.jpg');  
image(A);  
B=rgb2hsv(A);  
figure  
for i=1:3,  
subplot(2,2,i)  
imagesc(B(:,:,i))  
end  
subplot(2,2,4)  
imagesc(A);
```



# Intensity



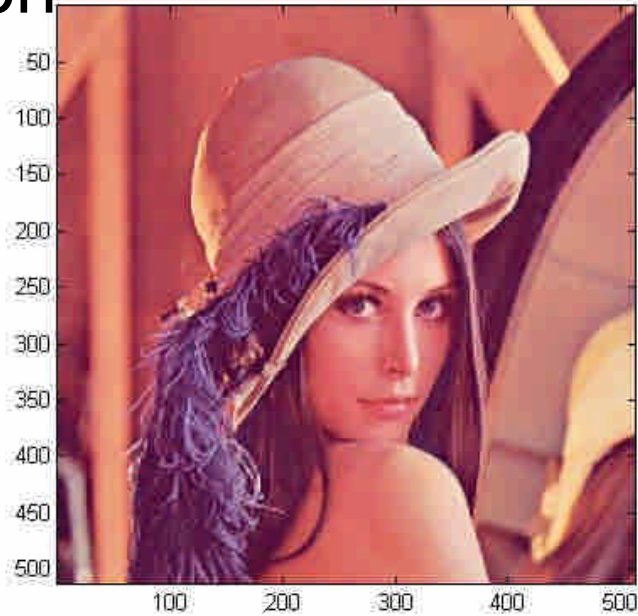
Original image



C

```
A=imread('lena.jpg');  
image(A);  
B=rgb2hsv(A);  
B1=B;  
B1(:,:,3)=B(:,:,3).^(1/2);  
C=hsv2rgb(B1);  
figure  
image(C)  
B1(:,:,3)=B(:,:,3).^(2);  
C=hsv2rgb(B1);  
figure  
image(C)
```

# Saturation



Original image



C



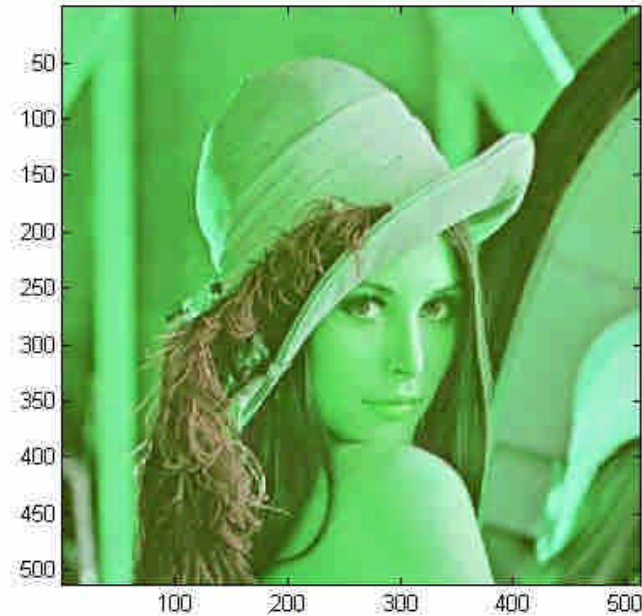
C

```
A=imread('lena.jpg');  
image(A);  
B=rgb2hsv(A);  
B1=B;  
B1(:,:,2)=B(:,:,2).^(1/2);  
C=hsv2rgb(B1);  
figure  
image(C)  
B1(:,:,2)=B(:,:,2).^(2);  
C=hsv2rgb(B1);  
figure  
image(C)
```

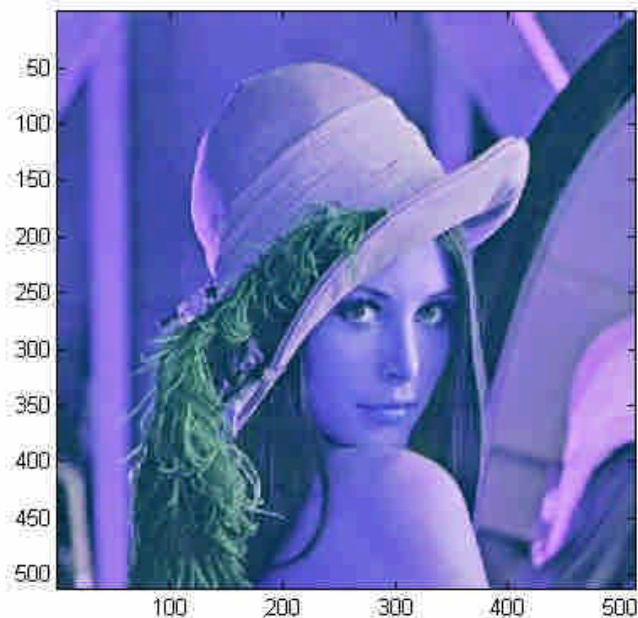
# Hue



Original image



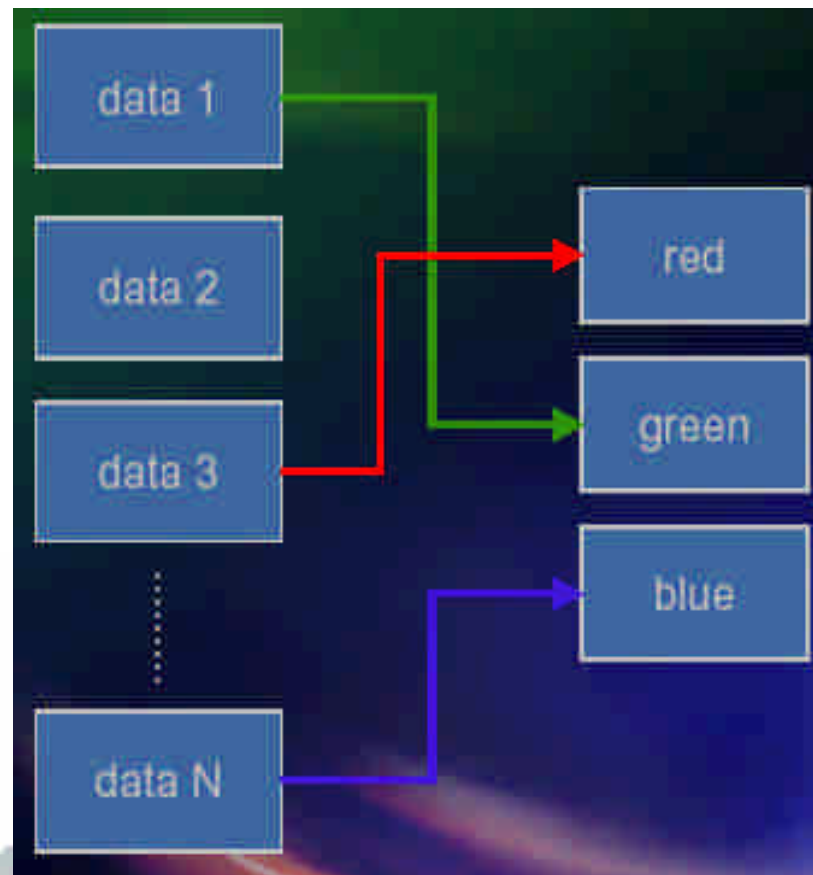
C



C

```
A=imread('lena.jpg');  
image(A);  
B=rgb2hsv(A);  
B1=B;  
B1(:,:,1)=mod(B(:,:,1)+(1/3),1);  
C=hsv2rgb(B1);  
image(C)  
B1(:,:,1)=mod(B(:,:,1)-(1/3),1);  
C=hsv2rgb(B1);  
image(C)
```

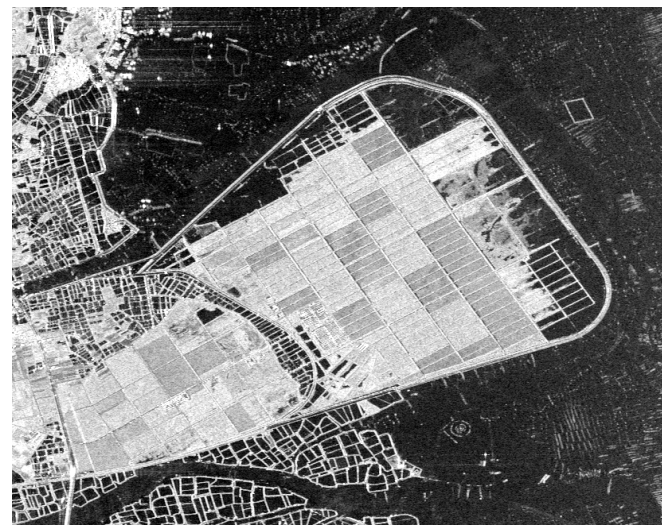
## 假色 (false color)



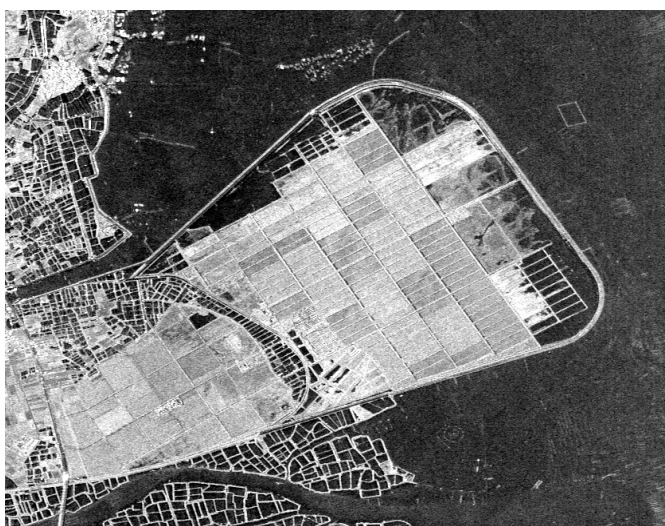
# SAR Image



(a) C-band |HV|



(b) C-band |HH|

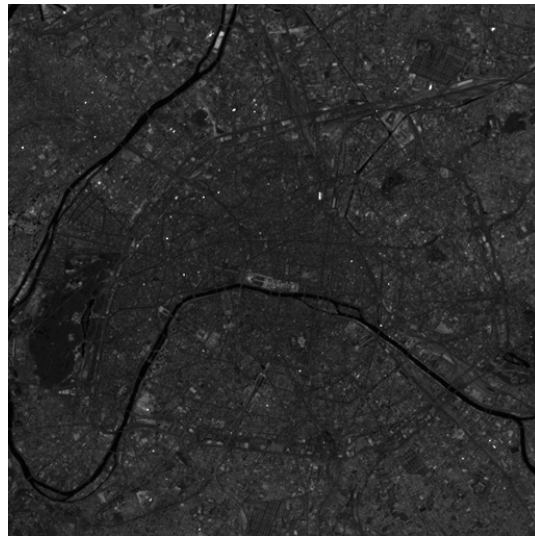


(c) C-band |VV|

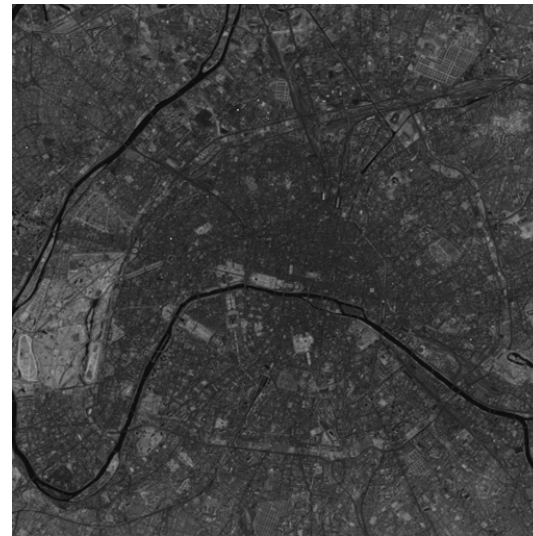


(d) 假色影像 (R: |HV| G: |HH| B: |VV|)

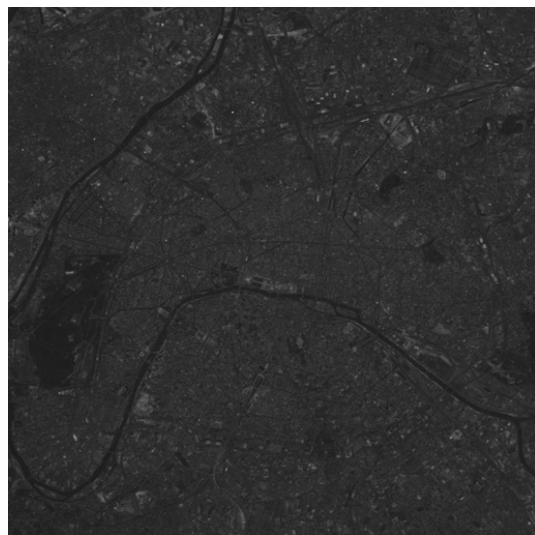
# Landsat Image



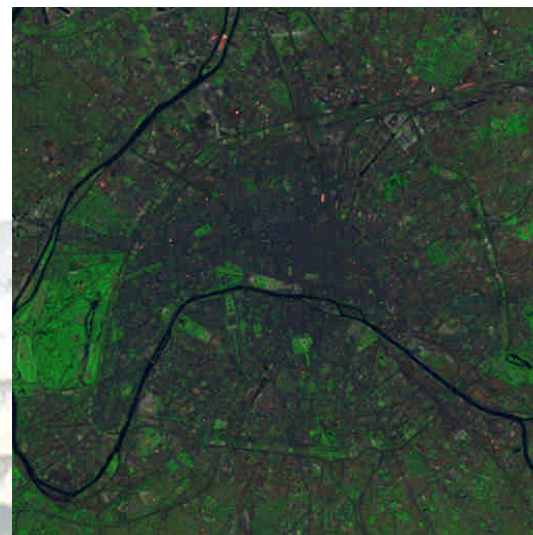
(a) band 7



(b) band 4

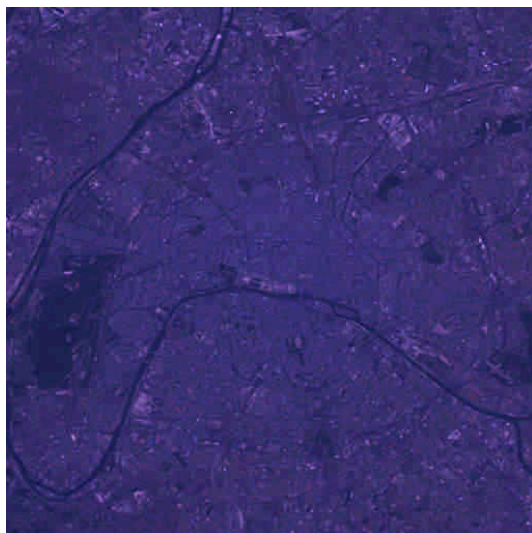


(c) band 2

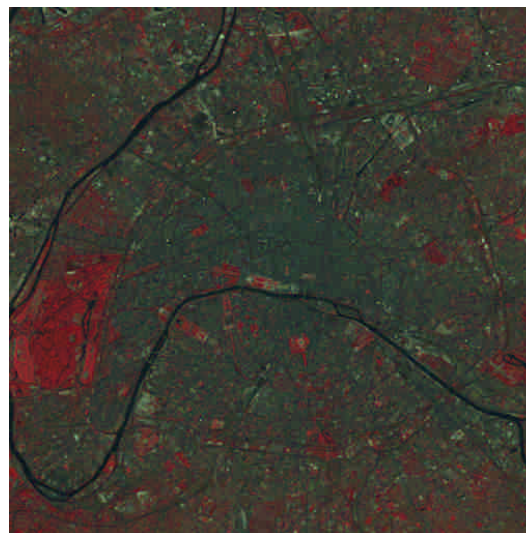


(d) 假色影像 (R: 7 G: 4 B: 2)

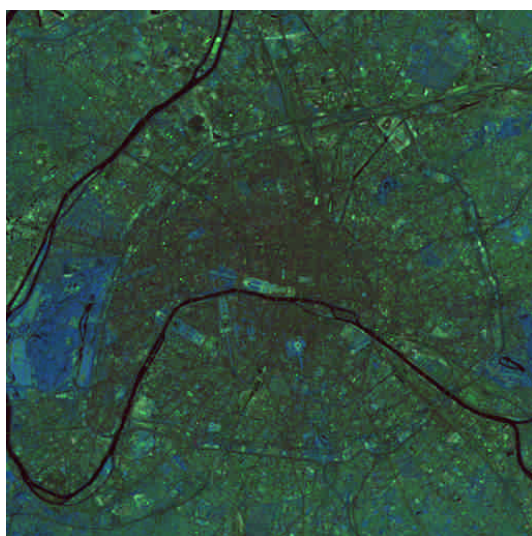
# Landsat (false color)



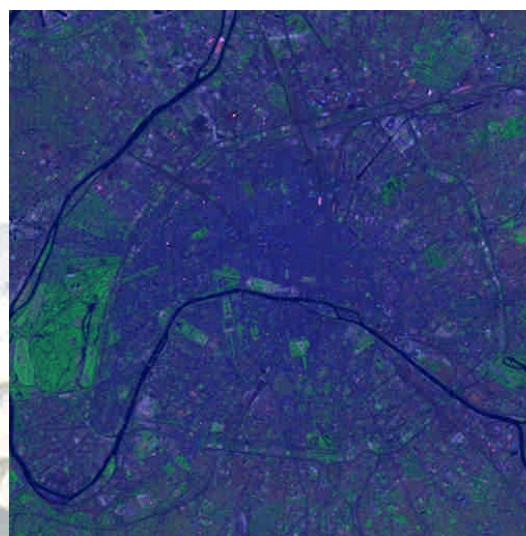
(c) 假色影像 (R: 3 G: 2 B: 1)



(d) 假色影像 (R: 4 G: 3 B: 2)

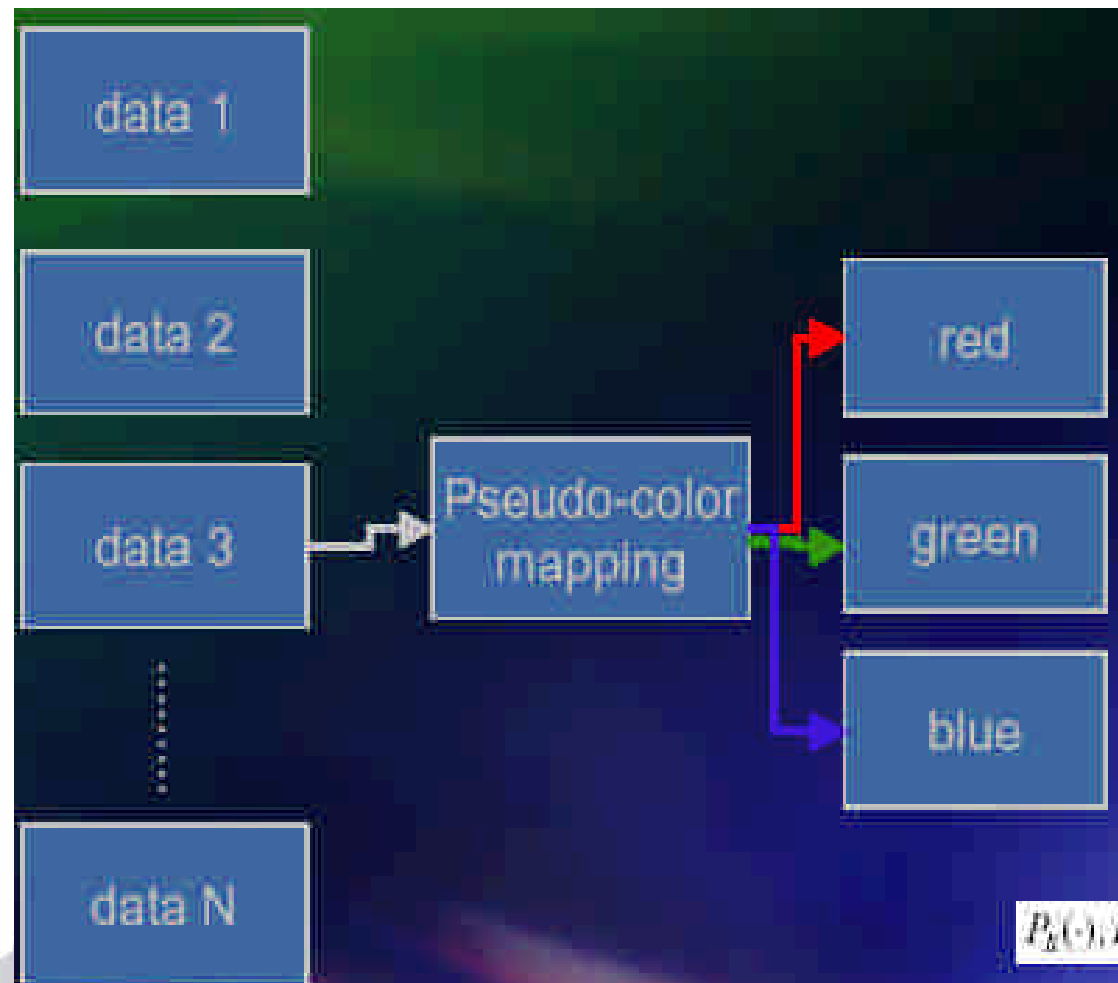


(c) 假色影像 (R: 2 G: 5 B: 4)

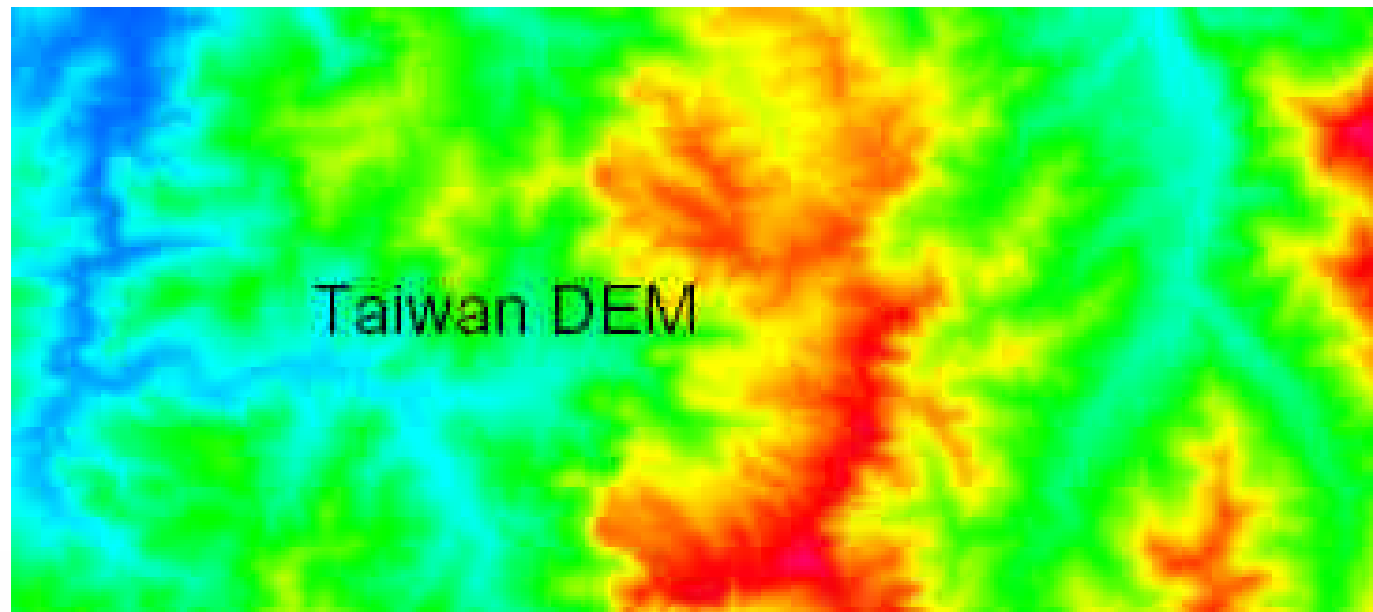


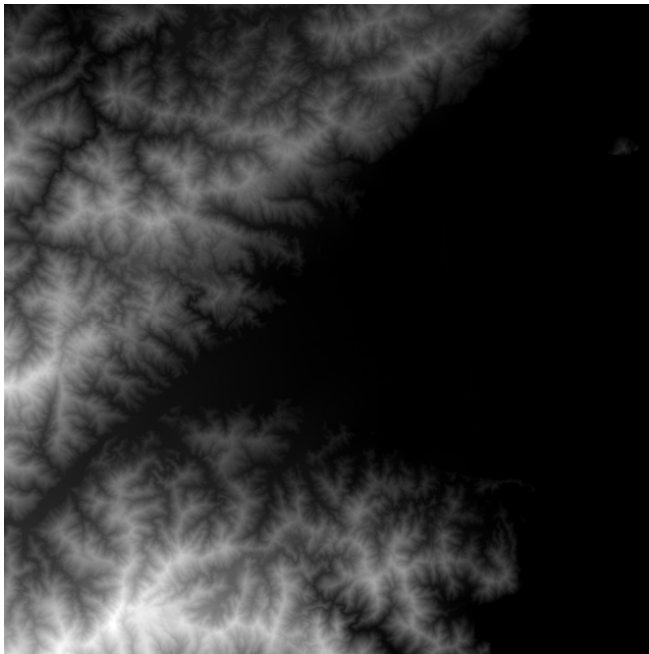
(d) 假色影像 (R: 7 G: 4 B: 1)

# 虛擬色 (Pseudo color)

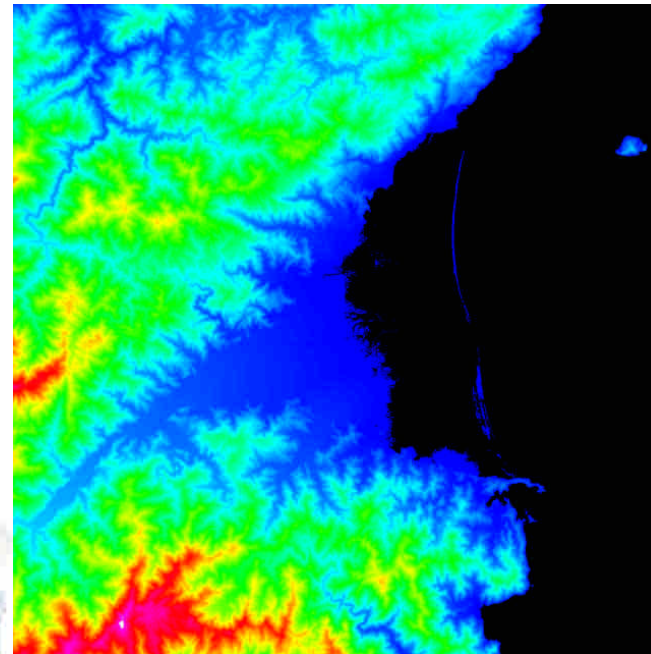


# Digital Elevation Model





(a) Gray level



(b) Pseudo color

# Color Space Transform

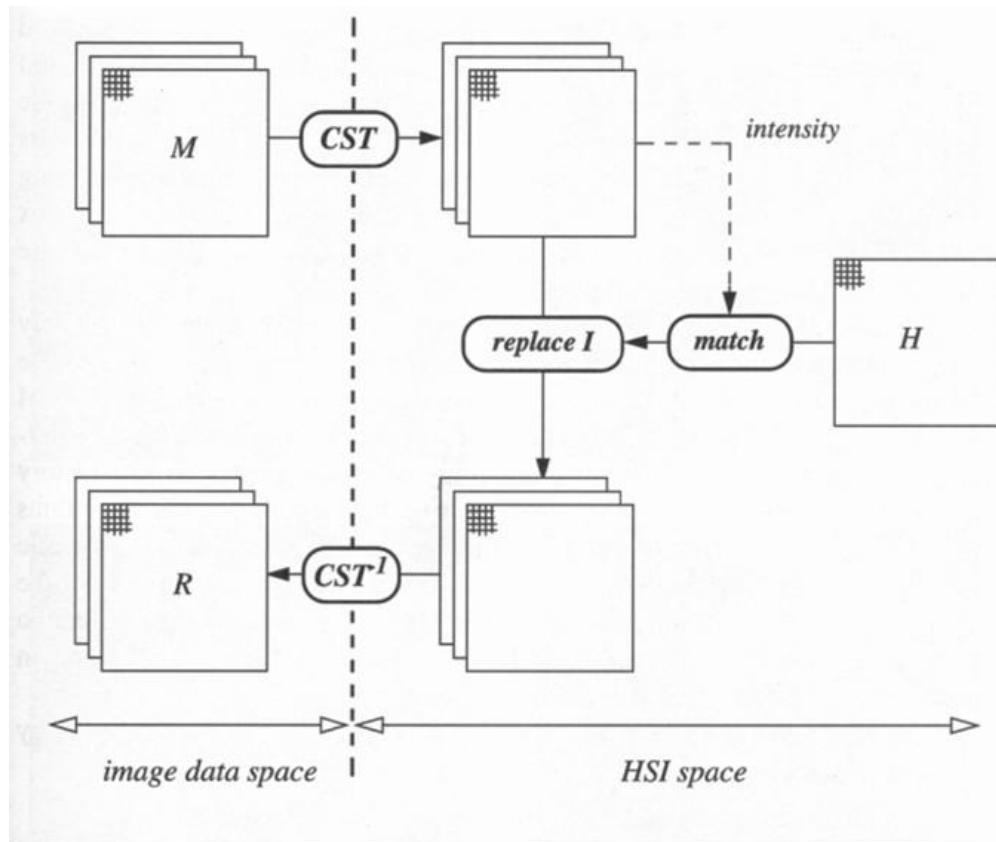
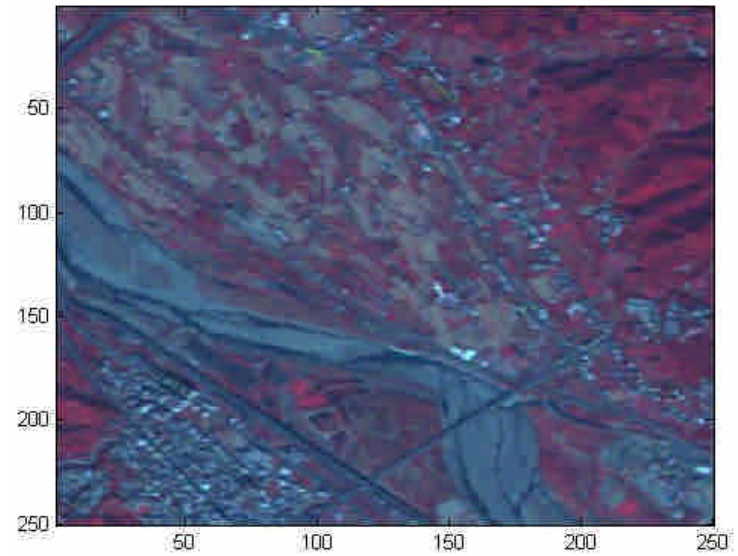
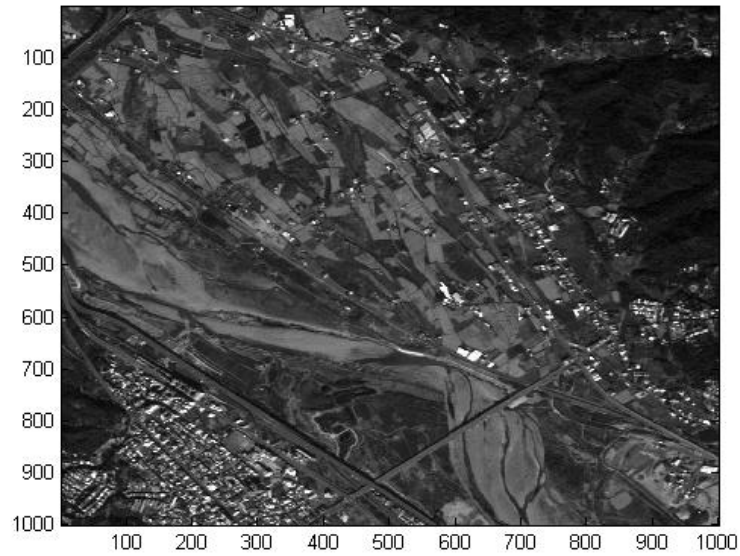


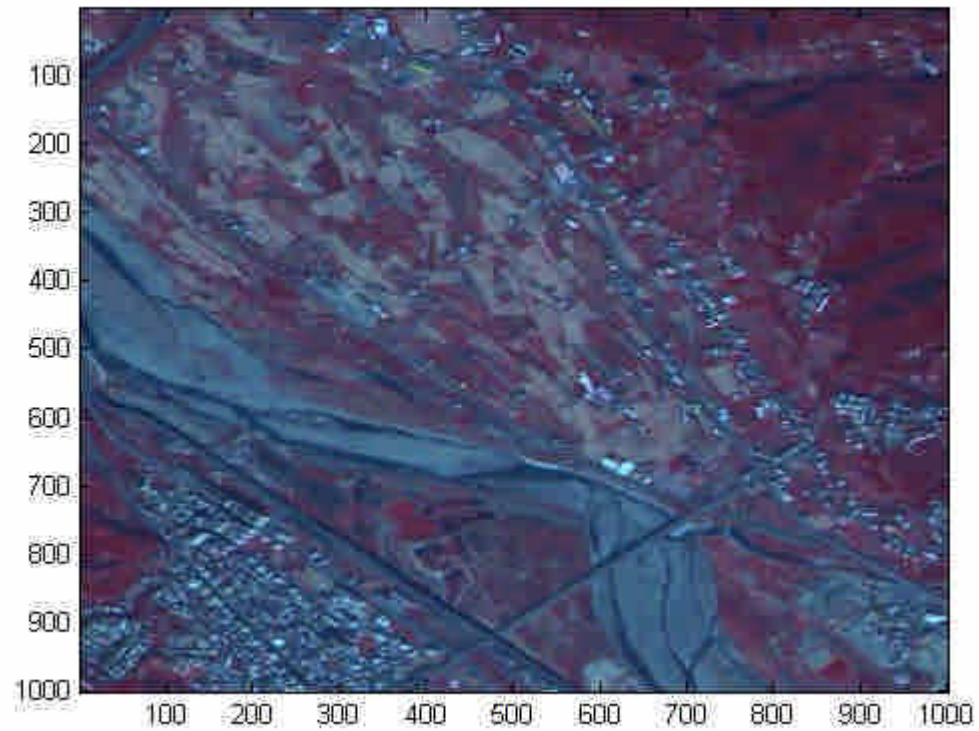
FIGURE 8-14. Image fusion using a color space transform (CST) and replacement of the intensity component. The multispectral image *M* has been previously registered to the high resolution image *H*.

# CST

```
load fusion_spot  
imagesc(A)  
colormap(gray)  
figure  
image(A1(:,:, [3 2 1]))
```

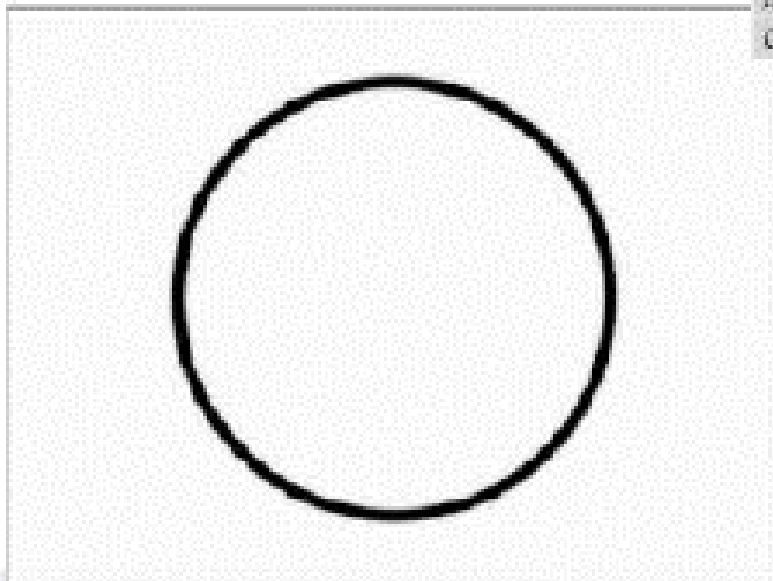
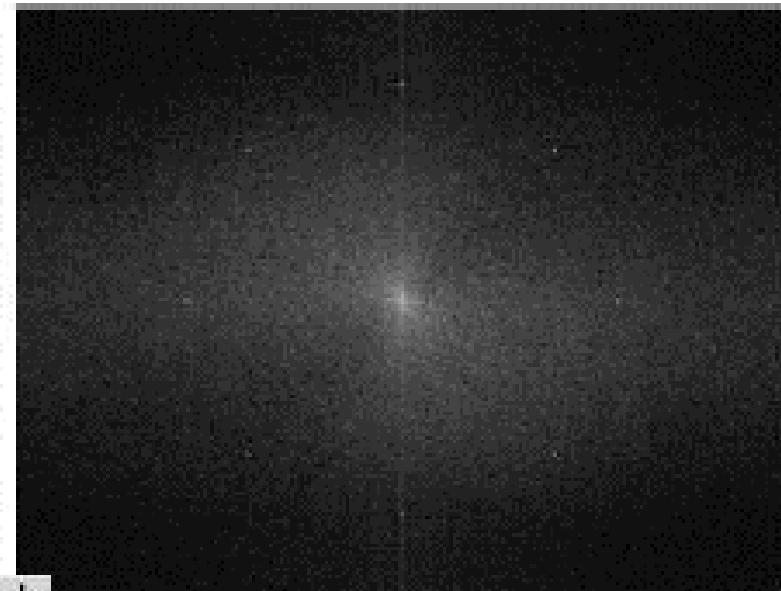
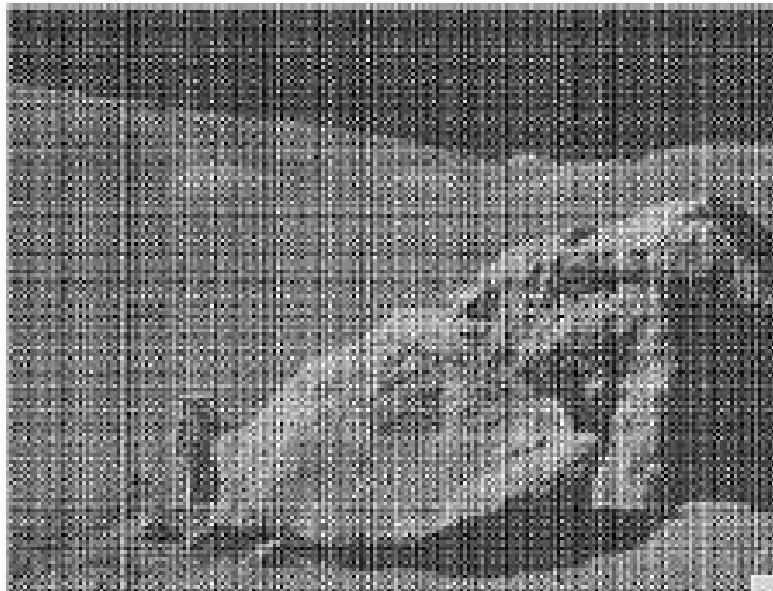


# CST



```
C=rgb2hsv(B);  
D=zeros(1000,1000,3);  
for i=1:4,for j=1:4  
    D(i:4:end,j:4:end,:)=C;  
end,end  
D(:,:,3)=double(A)/255;  
F=hsv2rgb(D);
```

# 特殊雜訊分布濾波器



a b  
c d

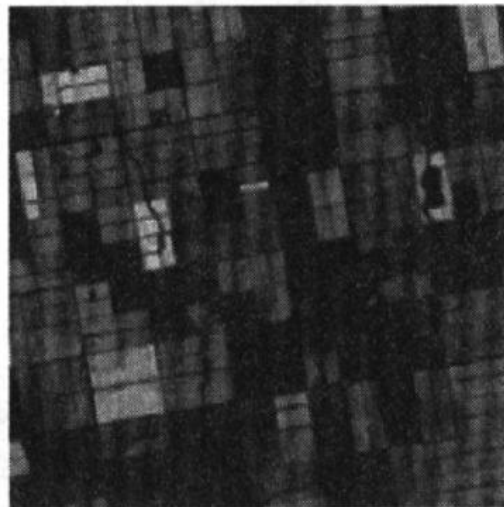
# 植生指數

Ratio Vegetation Index (RVI)

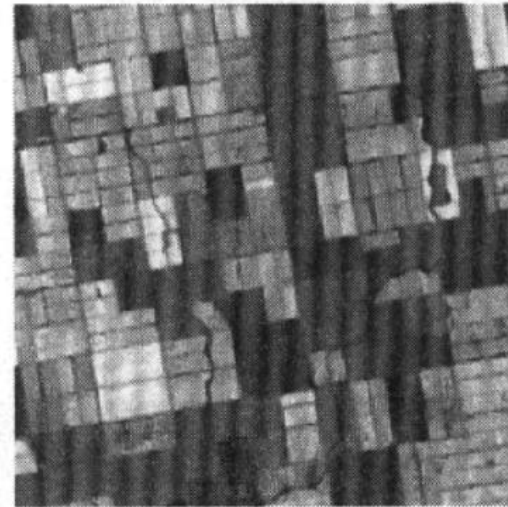
$$RVI = \frac{NIR}{R}$$

Normalized Difference Vegetation Index (NDVI)

$$NDVI = \frac{NIR - R}{NIR + R}$$



*RVI*



*NDVI*

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*FIGURE 5-5. The RVI and NDVI index images for the TM agriculture image of Plate 5-1. Bands 4 and 3 were used and the indexes are calculated from uncalibrated DN's.*

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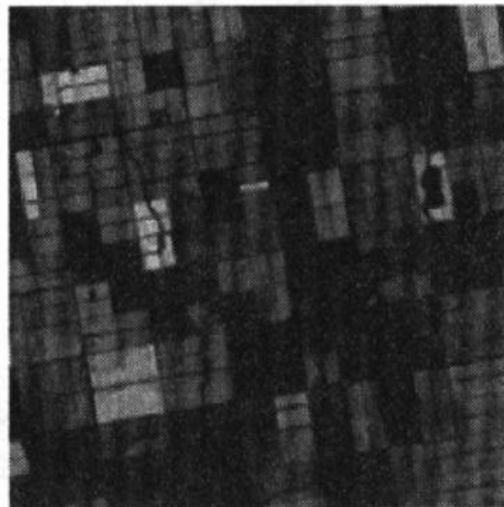
# 植生指數

Ratio Vegetation Index (RVI)

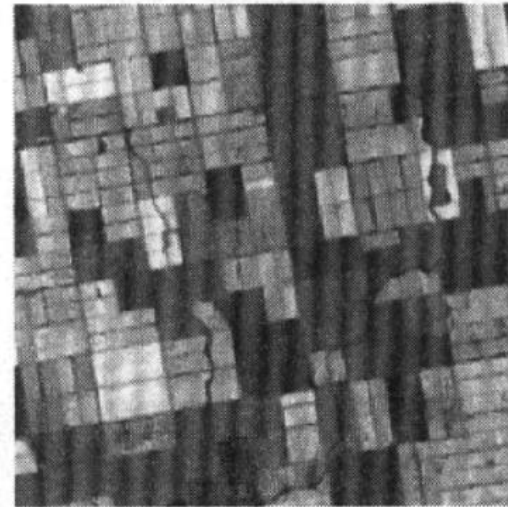
$$RVI = \frac{NIR}{R}$$

Normalized Difference Vegetation Index (NDVI)

$$NDVI = \frac{NIR - R}{NIR + R}$$



*RVI*



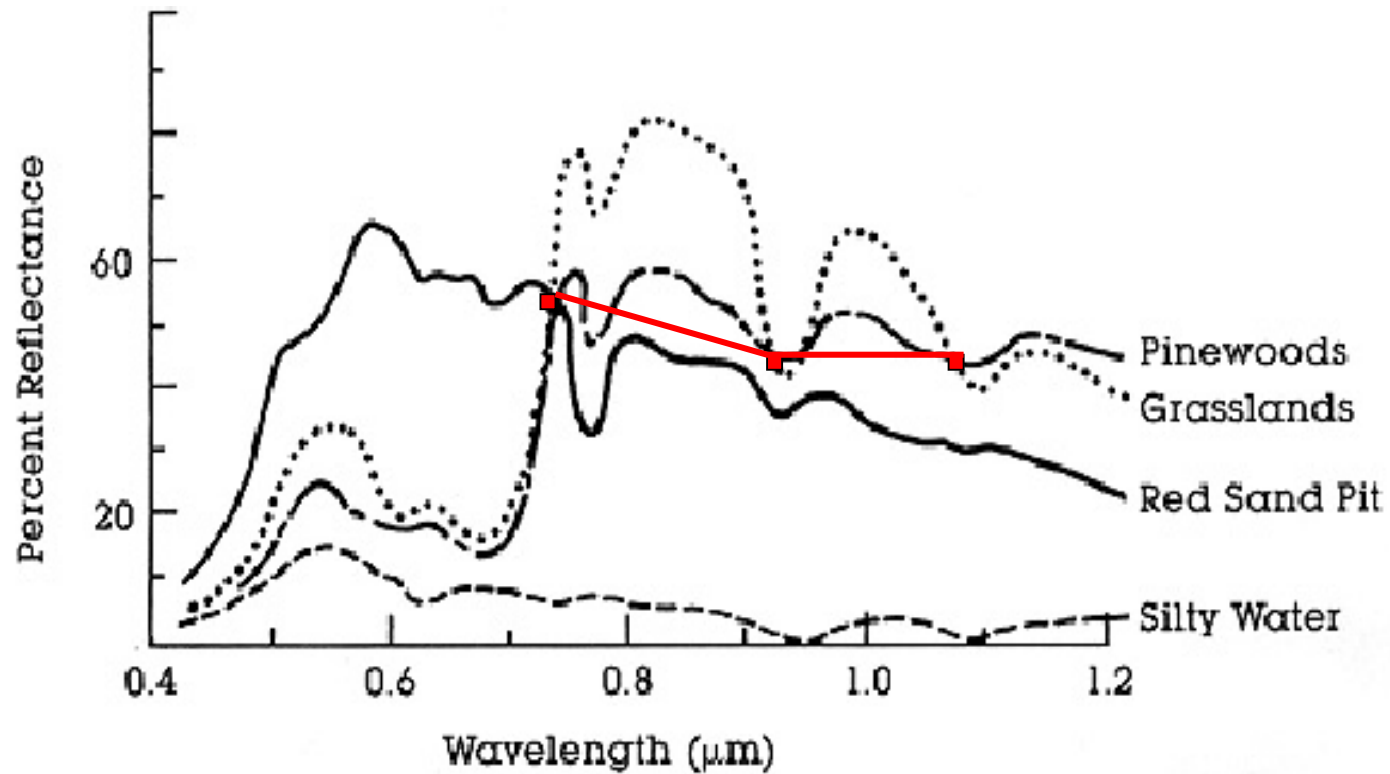
*NDVI*

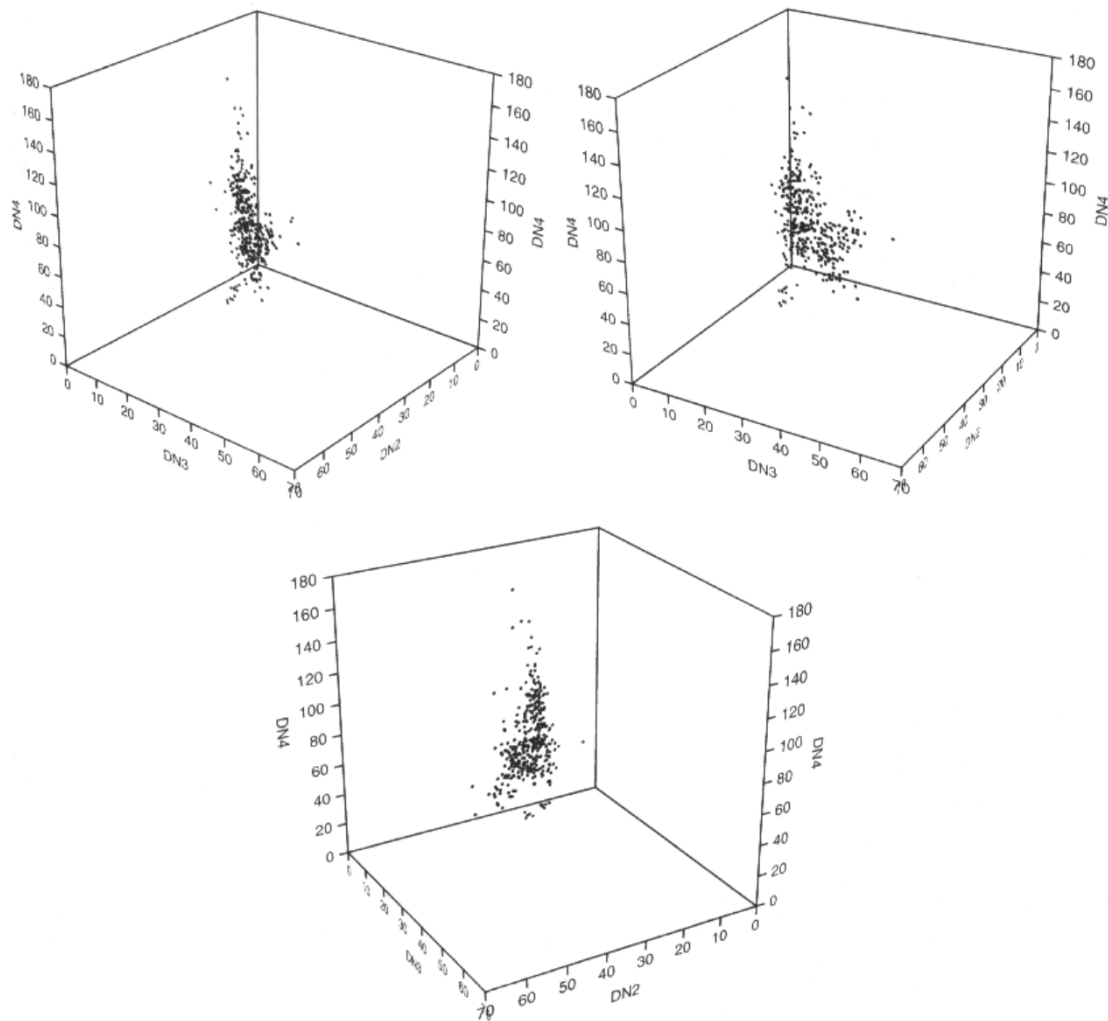
---

*FIGURE 5-5. The RVI and NDVI index images for the TM agriculture image of Plate 5-1. Bands 4 and 3 were used and the indexes are calculated from uncalibrated DN's.*

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# 頻譜解析度

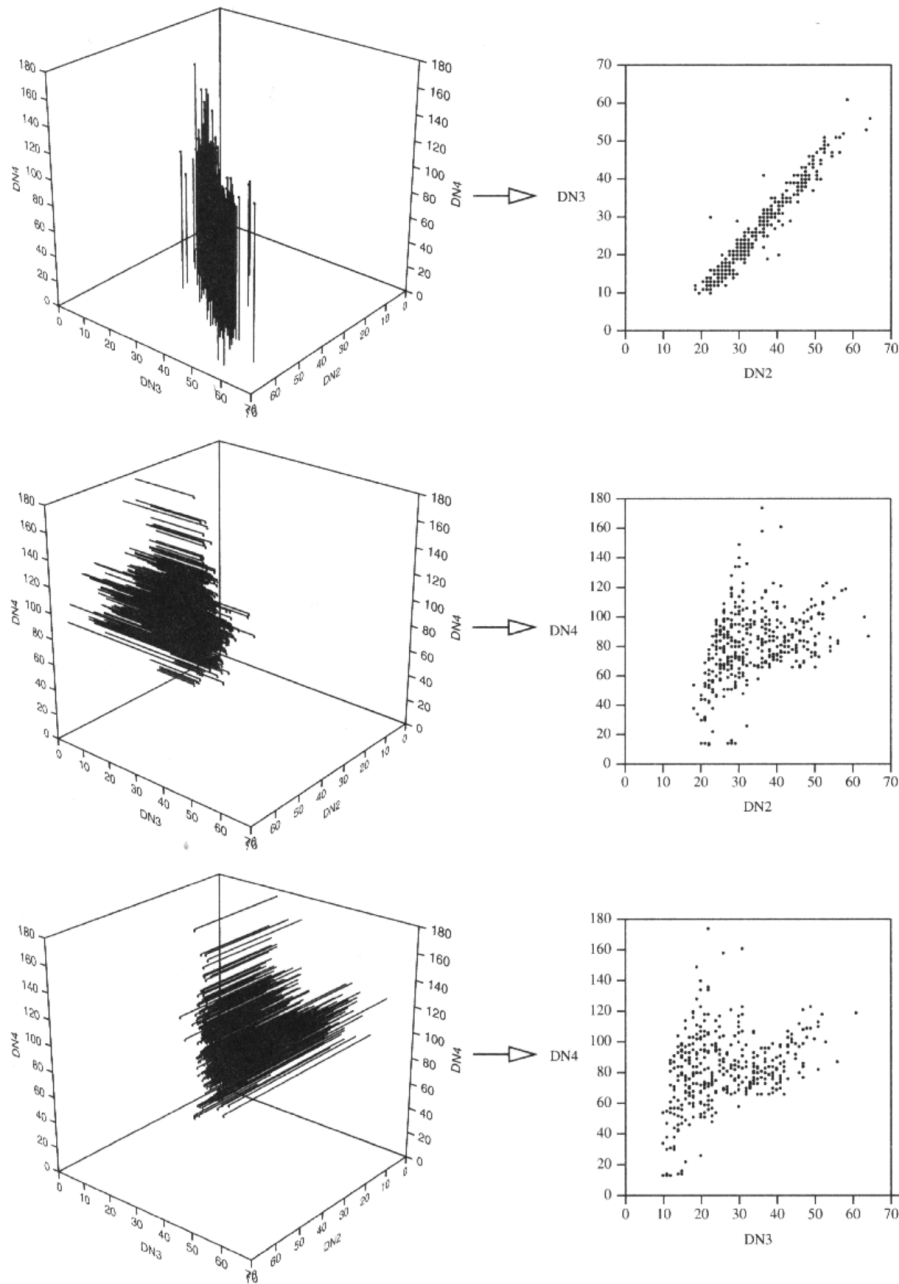




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*FIGURE 4-6. Three-band scatterplots of bands 2, 3 and 4 of a TM image, viewed from three different directions. Only every 20<sup>th</sup> sample and line of the image are used to calculate these scatterplots, so that they are not too dense. Every dot in the scatterplot represents one or more pixels with a particular spectral vector. Note that the image data occupies a small fraction of the total DN volume.*

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*FIGURE 4-7. Reduction of 3-D scatterplots to 2-D scatterplots by projections onto the three bounding planes. The 2-D scatterplots provide multiple views of the data, but do not contain all the information that exists in 3-D.*

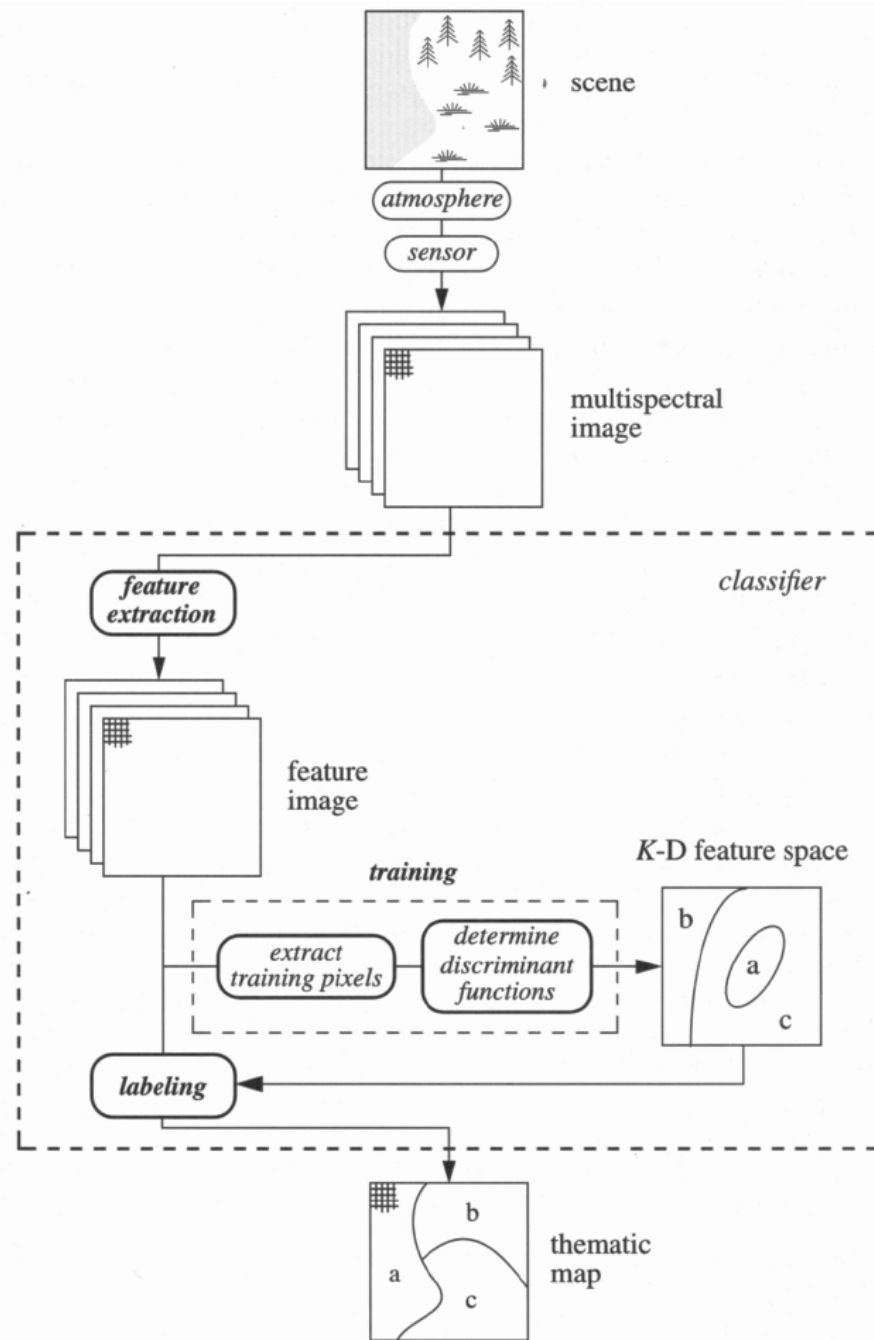
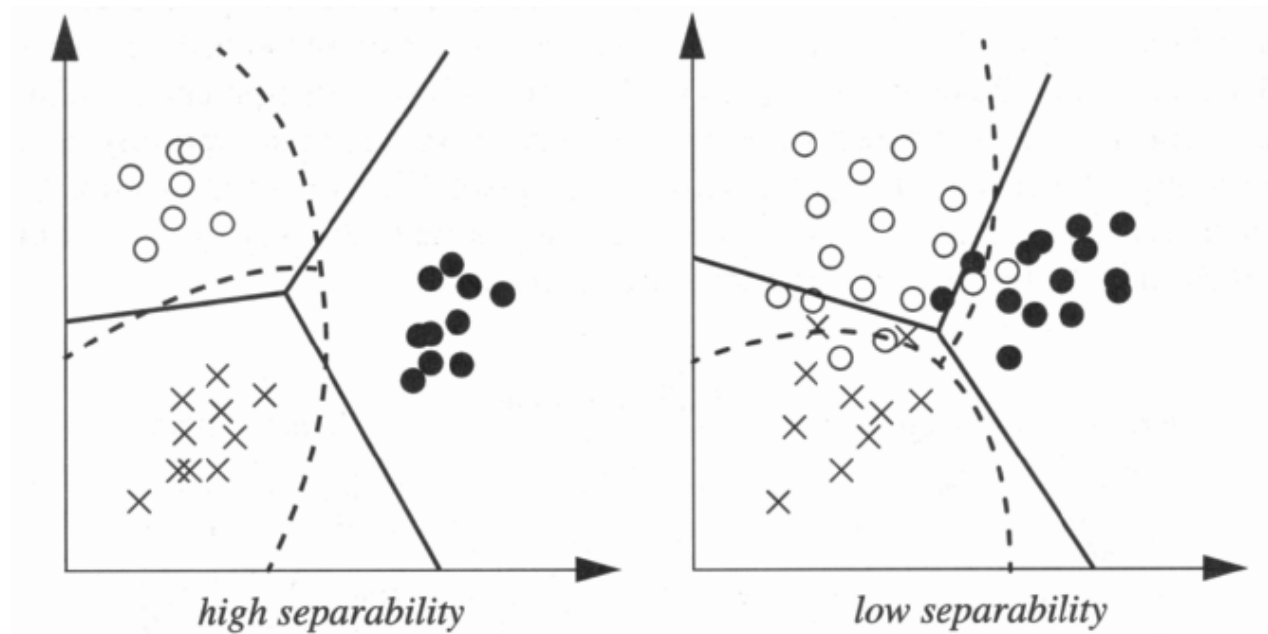


FIGURE 9-1. The data flow in a classification process.



*FIGURE 9-3. Two possible situations for training data in feature space and candidate decision boundaries. If the training classes are highly separable, there are many potential decision region partitions that can separate the classes without error; either the solid or dashed lines would do equally well. If the training data from different classes overlap, then a smooth decision boundary is impossible without misclassifications. The exact form of the decision boundary is then critical to the classification error.*

假設我們的資料點是由維度為 $m$ 的向量 $X_i$ 來表，  
其中  $i = 1, 2, \dots, n$

假設這一組資料的平均值是零：

$$\sum_{i=1}^n X_i = 0$$

目標是要找一個單位向量 $\mathbf{u}$ ，使得在 $\mathbf{u}$ 方向的投影平方和為最大

用一個  $m \times n$  矩陣 $\mathbf{A}$ 來表示此資料

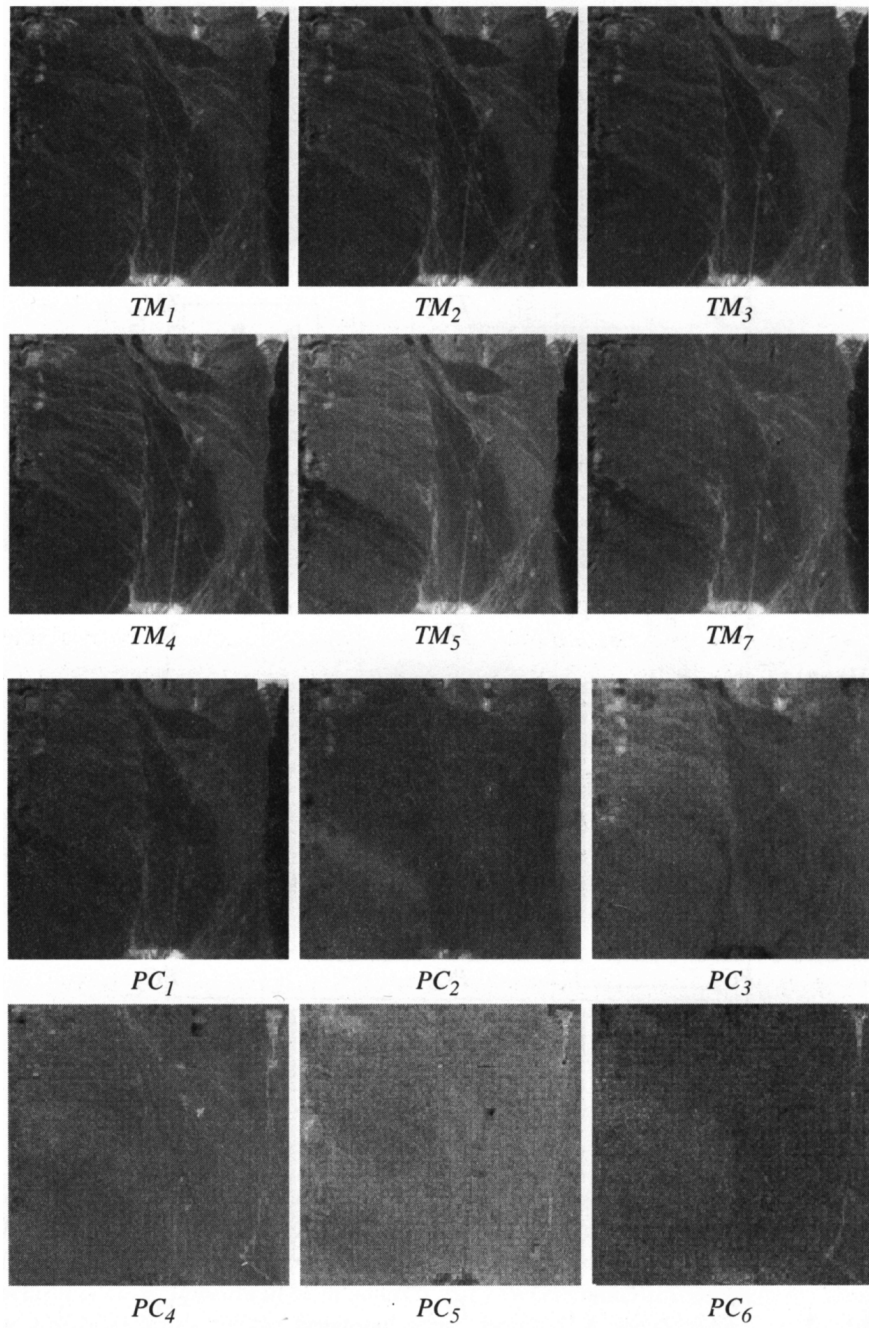
$$\mathbf{A} = \begin{bmatrix} \mathbf{X}_1 & \mathbf{X}_2 & \mathbf{X}_3 & \cdots & \mathbf{X}_n \end{bmatrix}$$

這一組資料在方向的投影可以表示為下列向量：

$$p = \begin{bmatrix} X_1^T u \\ X_2^T u \\ \vdots \\ X_n^T u \end{bmatrix} = A^T u$$

投影平方的總和為：

$$J(u) = \|p\|^2 = p^T p = (A^T u)^T (A^T u) = u^T A A^T u$$

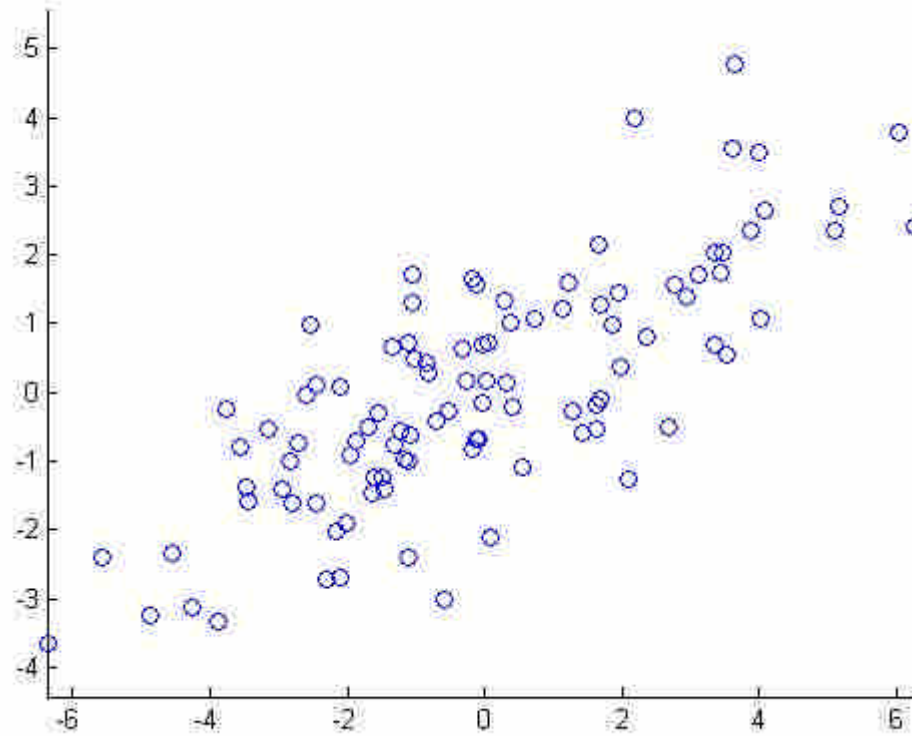


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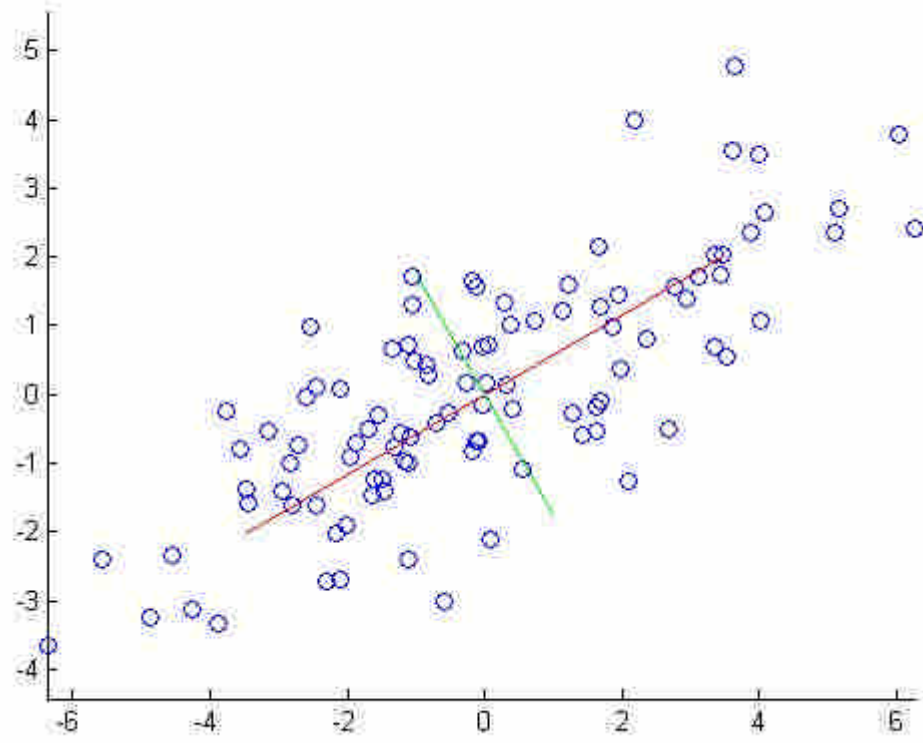
*FIGURE 5-13. PC transformation of a nonvegetated TM scene.*

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# Example



原始資料



兩各主軸方向