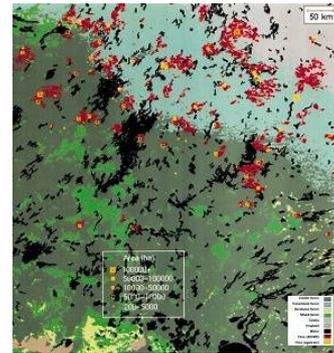


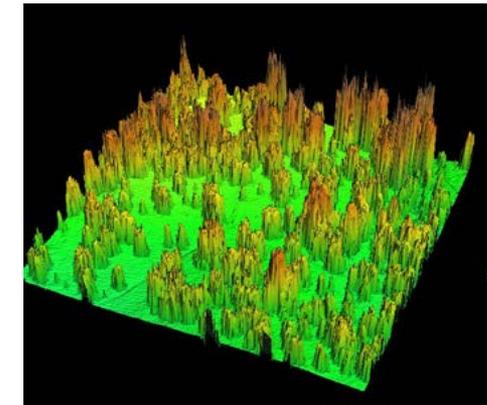
Processing of Remotely Sensed Data: from a Bunch of Numbers to...



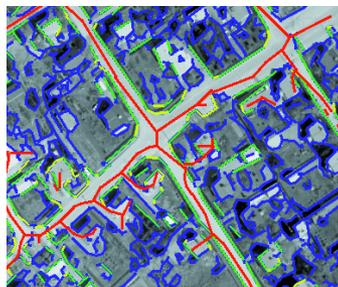
Land Cover Classification



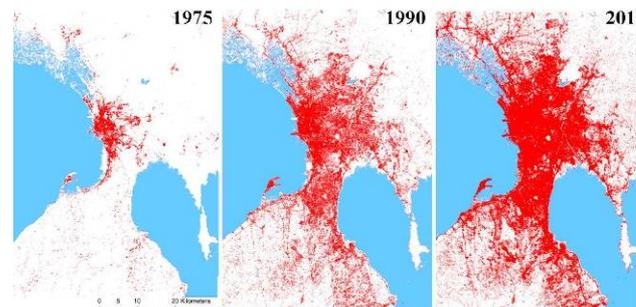
Target Detection
(here: Forest Fires)



3D Surface Modelling

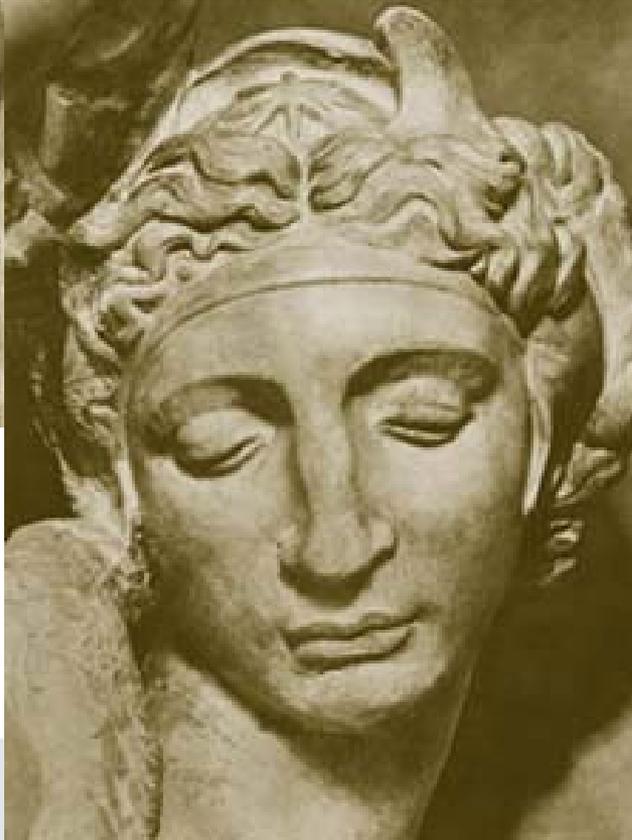


Feature Extraction



Multitemporal Analysis
(here: Urban Sprawl Monitoring)

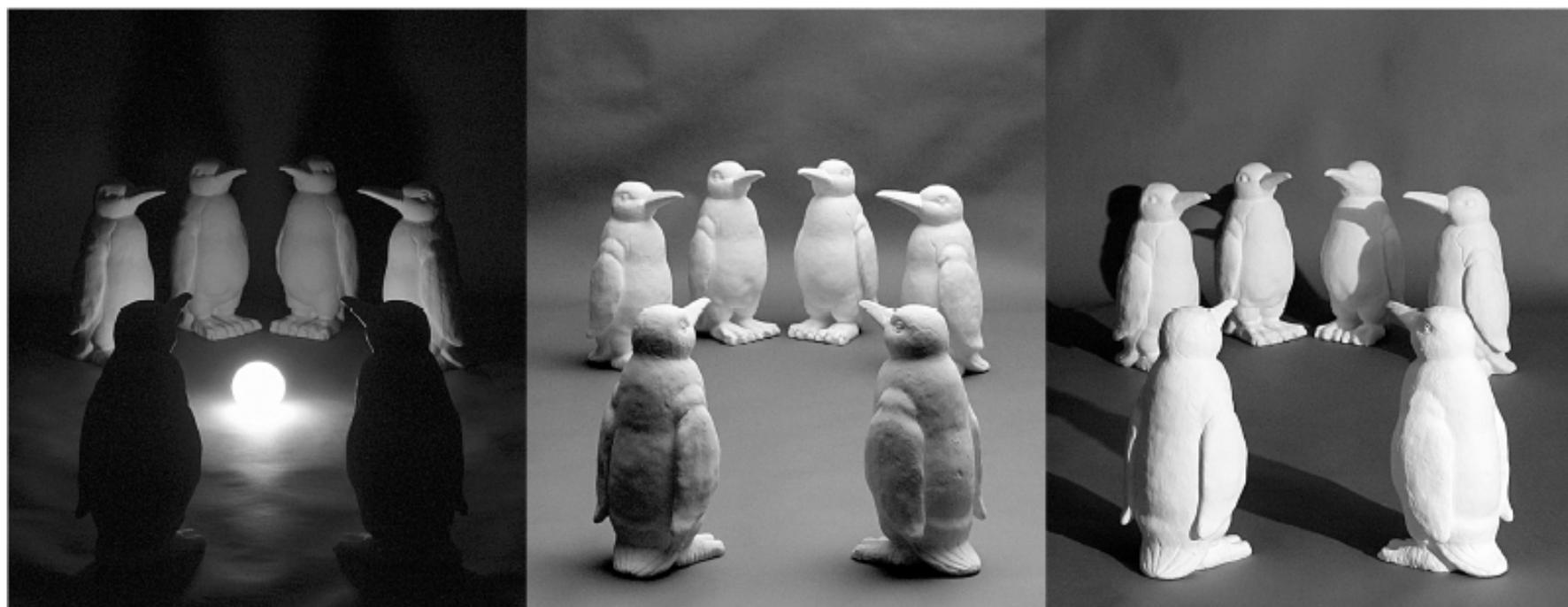
Challenges 1: view point variation



Michelangelo 1475-1564



Challenges 2: illumination



Challenges 3: occlusion

Magritte, 1957



Adapted from L. Fei-Fei,
R. Fergus, A. Torralba



Deutsches Zentrum
für Luft- und Raumfahrt e.V.
in der Helmholtz-Gemeinschaft

Challenges 4: scale



Challenges 5: background clutter

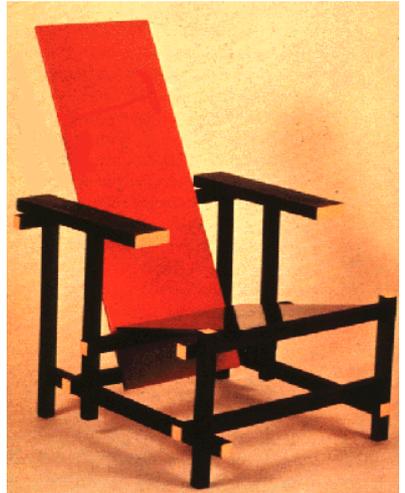


Dragon shrimp and commercial crab on a sea cucumber in Fiji.
Photograph by Tim Laman

NATIONAL GEOGRAPHIC

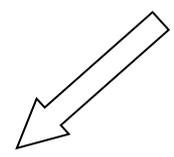
© 2007 National Geographic Society. All rights reserved.

Challenges 6: intra-class variation

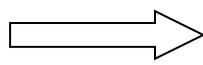
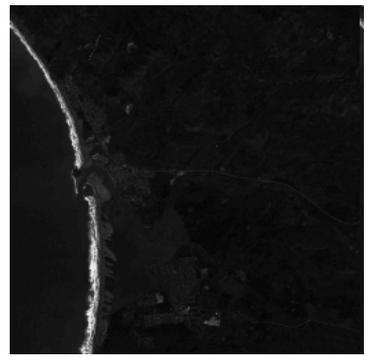


- Image Acquisition & Correction
 - Raw Data → Raw Image → Image

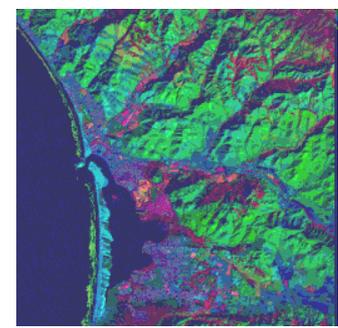
03	29	38	48
59	96	94	04
05	06	96	97
87	76	75	45



- Low-level Analysis
 - Image → Image
 - Time domain
 - Frequency domain



- Mid-level Analysis
 - Image → Features / Attributes
 - Feature Extraction
 - Clustering / Segmentation



- High-level Analysis
 - Features → Recognition
 - Classification

	Beach Bar		Urban Area
	Wave Breakers		Shadows
	Vegetation1		Sea
	Vegetation2		Mountains (bright slopes)
	Golf Course		

A nice book!

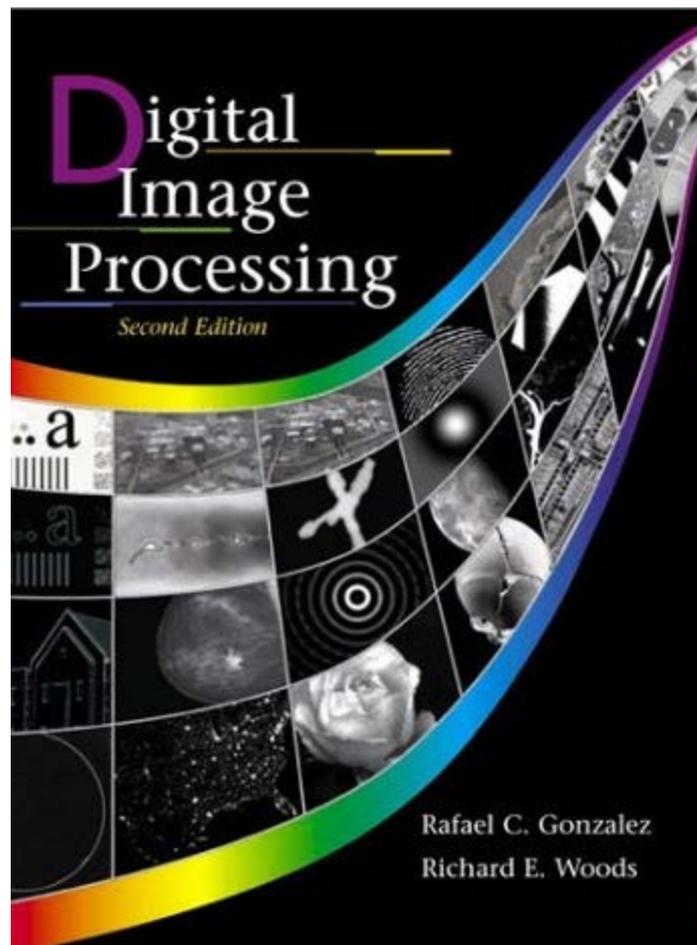
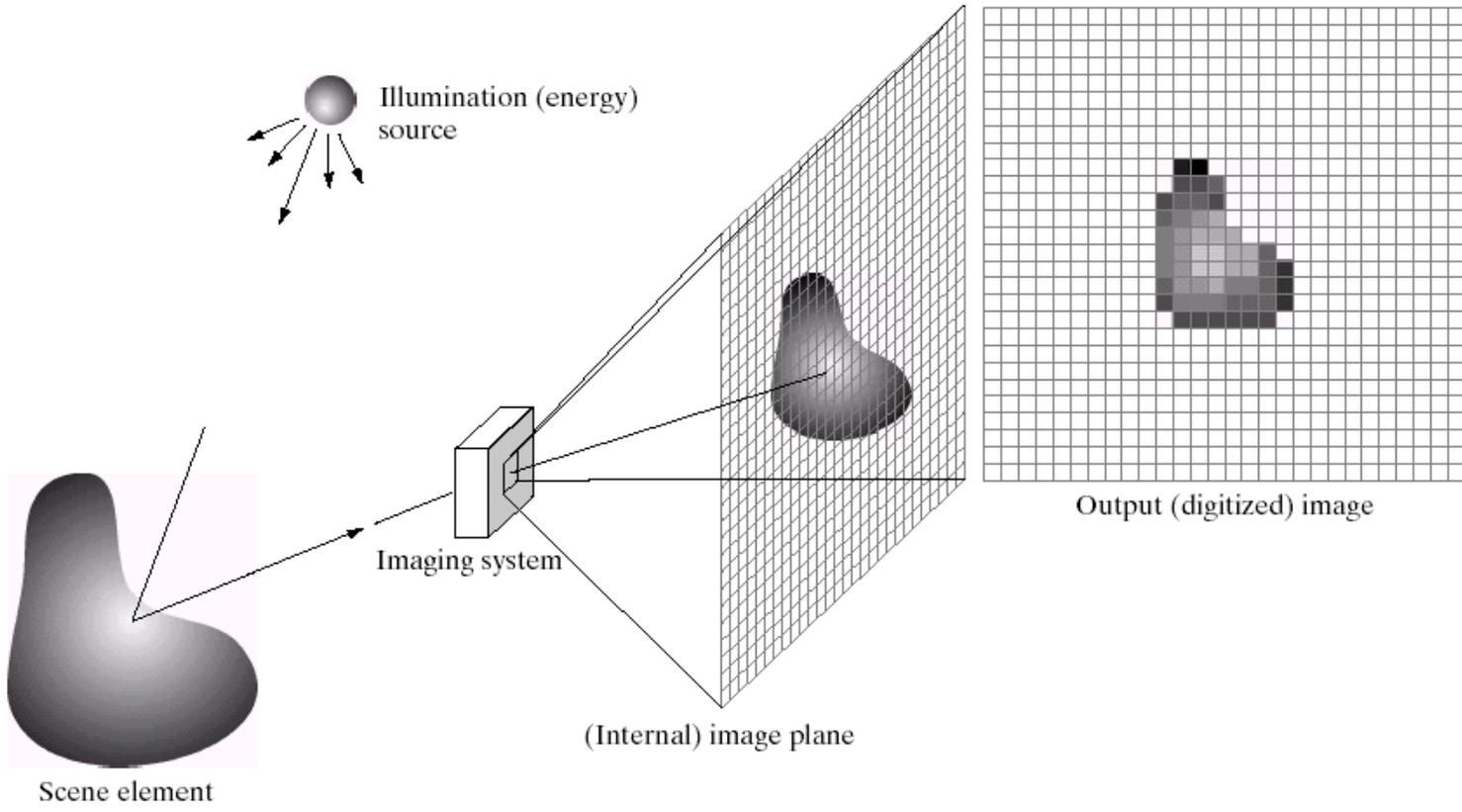


Image Acquisition



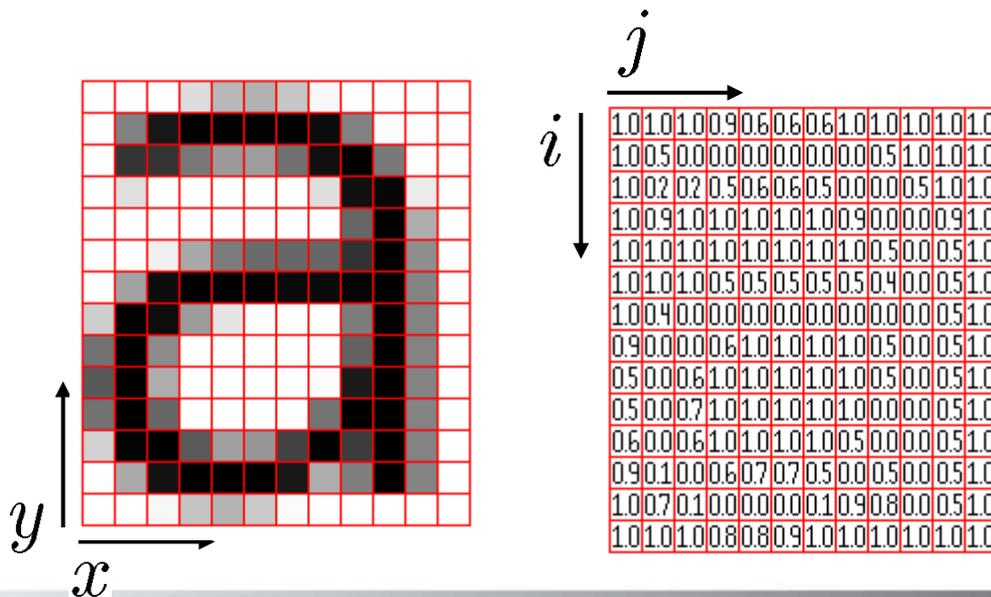
a b c d e

FIGURE 2.15 An example of the digital image acquisition process. (a) Energy (“illumination”) source. (b) An element of a scene. (c) Imaging system. (d) Projection of the scene onto the image plane. (e) Digitized image.

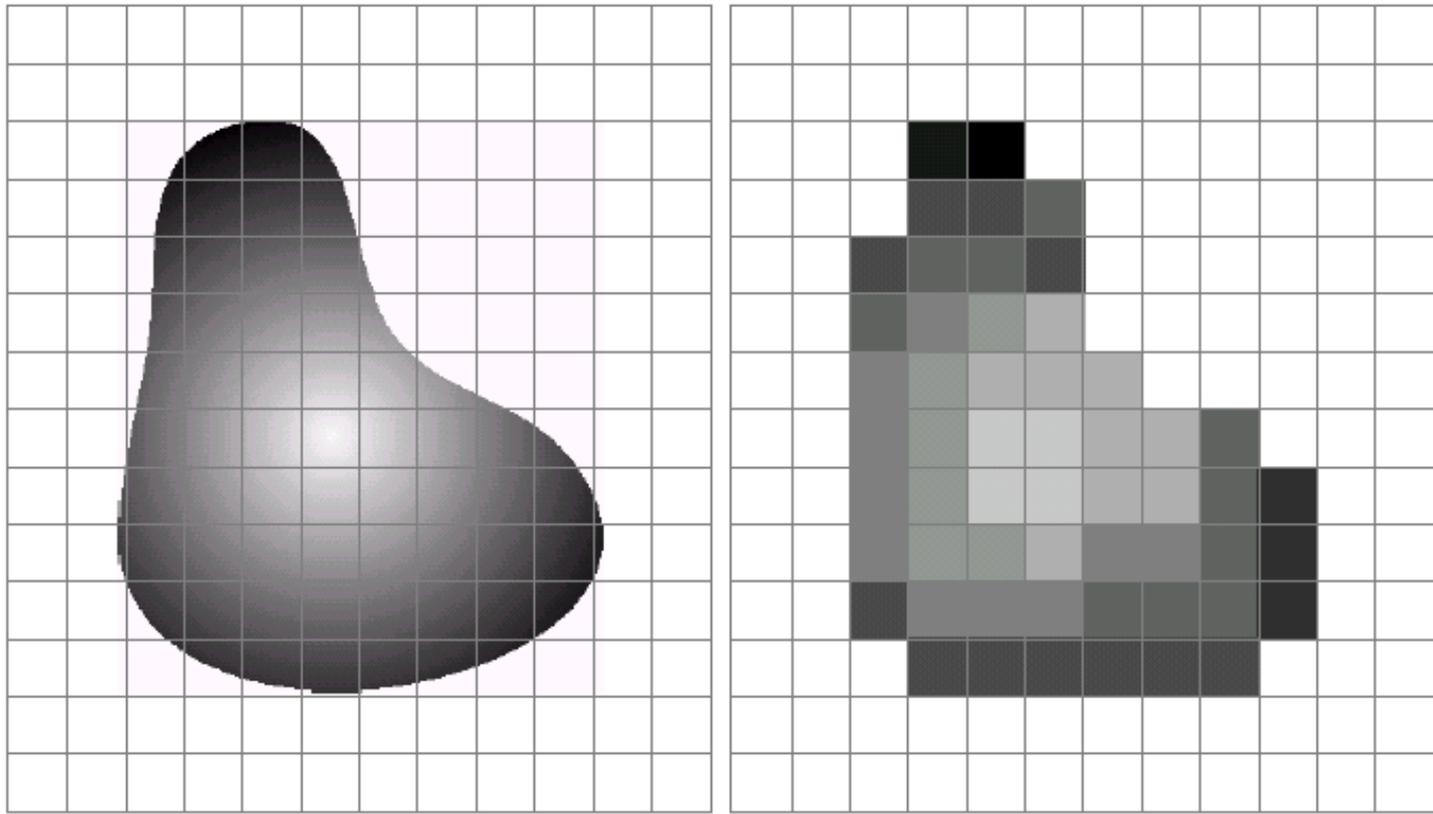
Image representation

- In a **digital** image, both the coordinates and the image value become **discrete** quantities
- Images can be represented as 2D arrays (matrices) of integer values: $I[i,j]$ (or $I[r,c]$)
- The term **gray level** is used to describe monochromatic intensity

A rasterized form of the letter 'a' magnified 16 times



Sampling and quantization



a b

FIGURE 2.17 (a) Continuous image projected onto a sensor array. (b) Result of image sampling and quantization.



Spatial resolution ← Sampling

- Spatial resolution is the smallest discernible detail in an image
- Sampling is the principal factor determining spatial resolution



1024



512



256



128



64



32



Spatial resolution



1024 x 1024



512 x 512



256 x 256



128 x 128



64 x 64



32 x 32

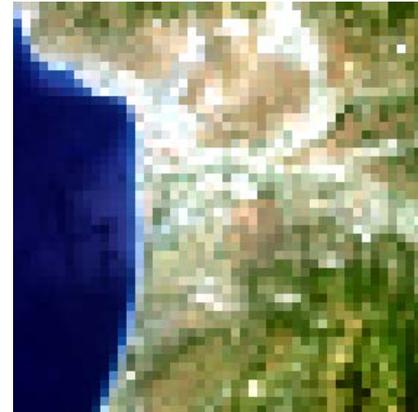


Spatial resolution: Resampling

Resampling without interpolation (nearest-neighbour resampling)



128 x 128



64 x 64



32 x 32

Resampling with interpolation (each pixel is a combination of neighbouring pixels)



128 x 128



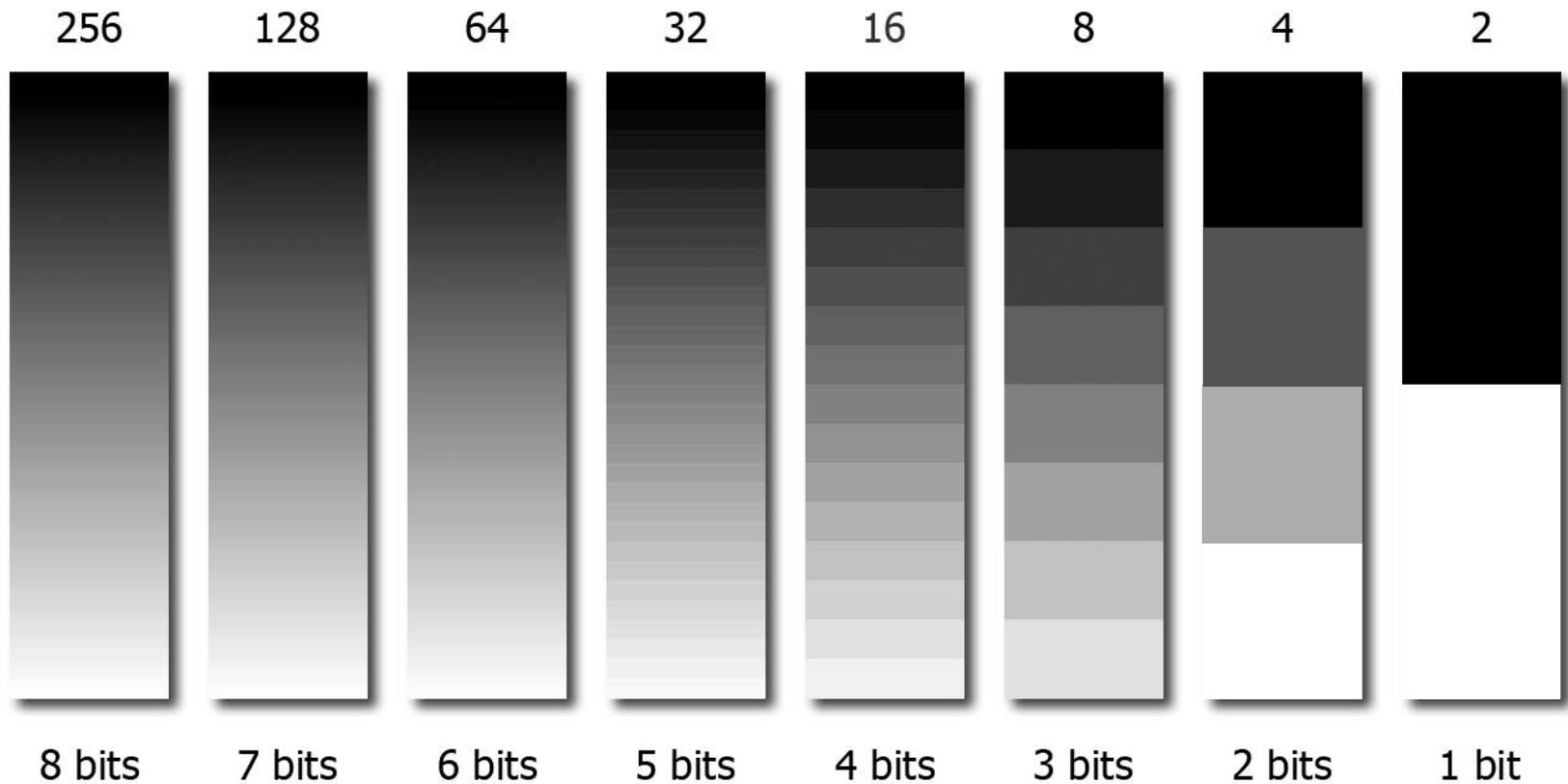
64 x 64



32 x 32

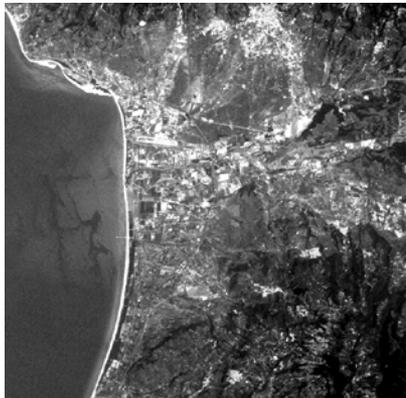


Radiometric Resolution



Radiometric resolution \leftarrow Quantization

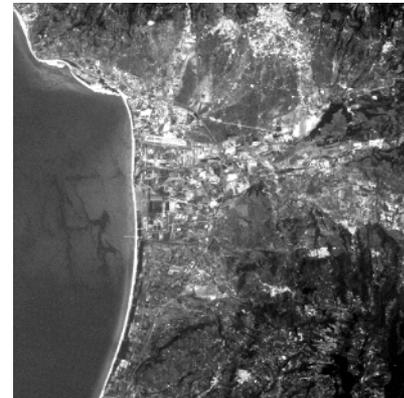
- Radiometric resolution refers to the smallest discernible change in gray level (often power of 2)
- The human eye is inefficient at distinguishing differences in gray levels much beyond the limit of 16 (but to the machine it may make a big difference)



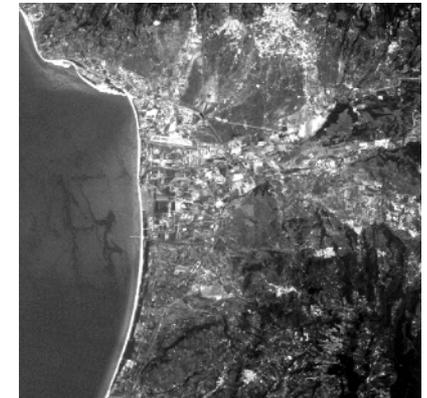
256 gray levels



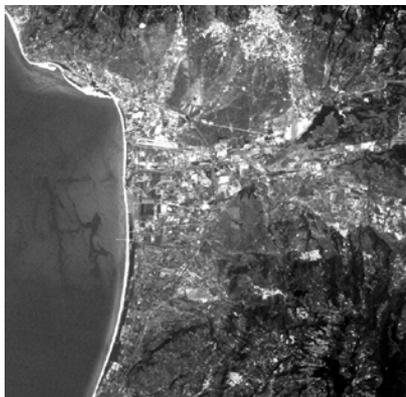
128



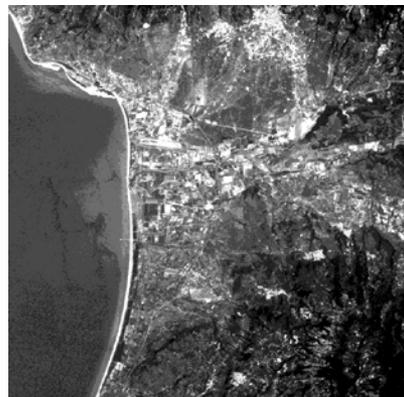
64



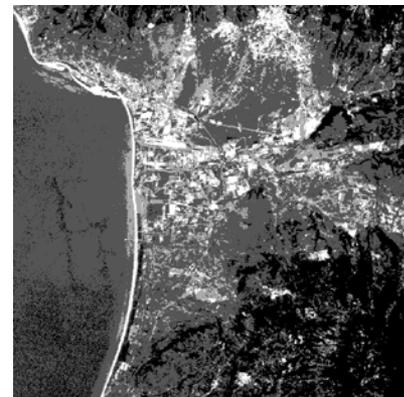
32



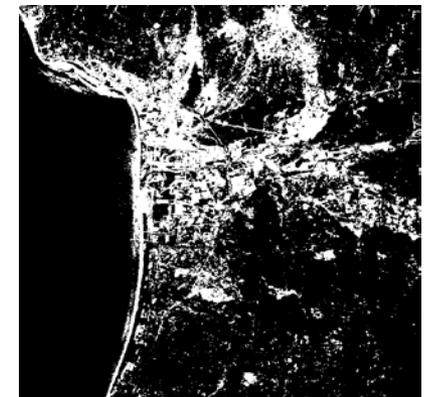
16



8

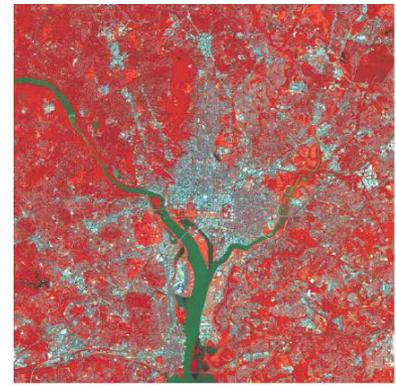
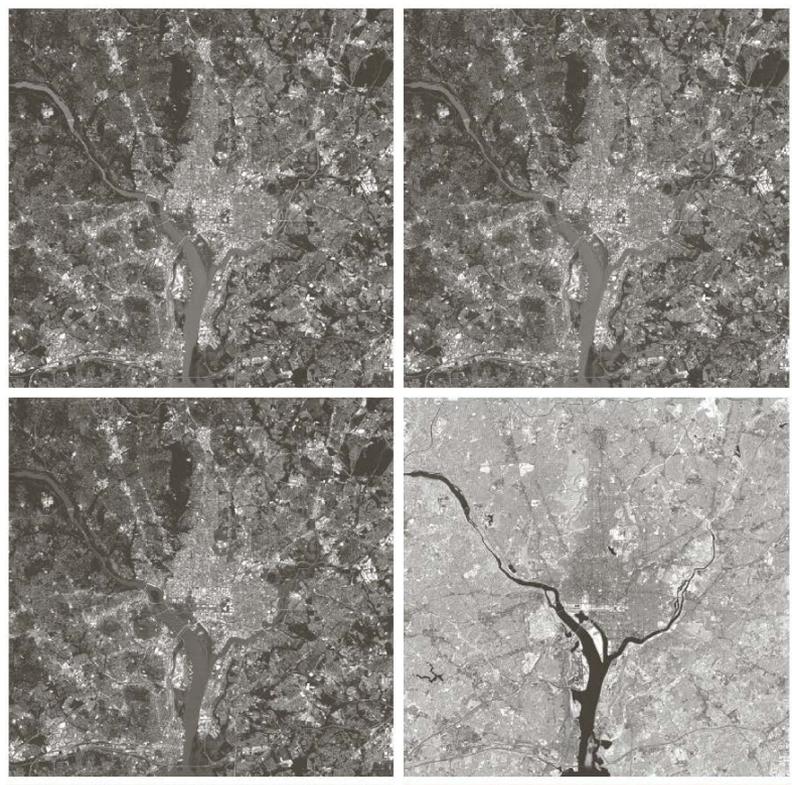


4



2

Spectral Resolution & Color Display



R G B
3 2 1

NIR R G
4 3 2

Reflected energy for each pixel in the frequencies Blue, Green, Red & Near Infrared

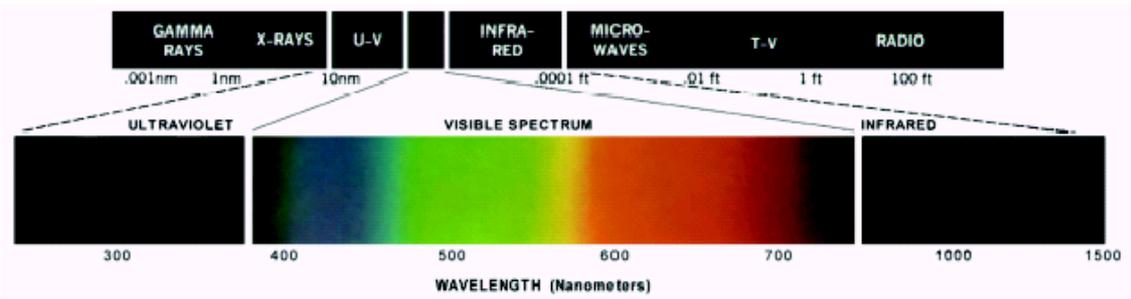


FIGURE 6.2 Wavelengths comprising the visible range of the electromagnetic spectrum. (Courtesy of the General Electric Co., Lamp Business Division.)

Bit planes

10000101



a	b	c
d	e	f
g	h	i

FIGURE 3.14 (a) An 8-bit gray-scale image of size 500×1192 pixels. (b) through (i) Bit planes 1 through 8, with bit plane 1 corresponding to the least significant bit. Each bit plane is a binary image.

Application of Quantization: Steganography

Pieter Bruegel (the Elder, ca. 1525-69), *The Peasant Dance*, 1568, Oil on oak panel, 114x164 cm, Kunsthistorisches Museum Wien, Vienna



10000101

If an image is quantized, say from 8 bits to 6 bits and redisplayed it can be all but impossible to tell the difference between the two.

6-bit-per-band, 3-band, quantized image

Application of Quantization: Steganography



Pieter Bruegel (the Elder, ca. 1525-69), *The Peasant Dance*, 1568, Oil on oak panel, 114x164 cm, Kunsthistorisches Museum Wien, Vienna

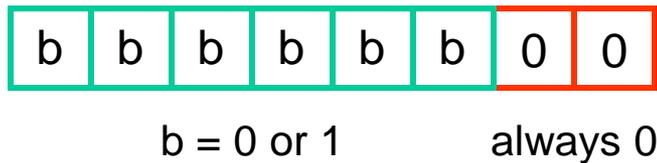
10000101

a value from the set {0, 4, 8, ... , 252}

If an image is quantized, say from 8 bits to 6 bits and redisplayed it can be all but impossible to tell the difference between the two.

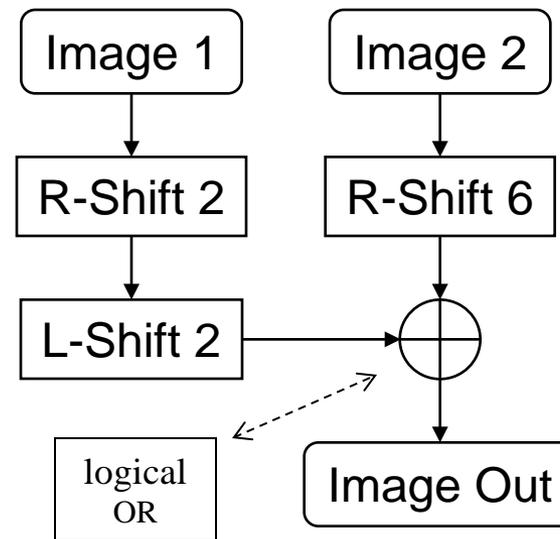
Application of Quantization: Steganography

If the 6-bit version is displayed as an 8-bit image then the 8-bit pixels all have zeros in the lower 2 bits:



This introduces the possibility of encoding other information in the low-order bits.

That other information could be a message, perhaps encrypted, or even another image.



X-Shift n = logical left or right shift by n bits.

Image 1 in upper 6-bits.
Image 2 in lower 2-bits.

Application of Quantization: Steganography



Pieter Bruegel (the Elder, ca. 1525-69), *The Peasant Dance*, 1568, Oil on oak panel, 114x164 cm, Kunsthistorisches Museum Wien, Vienna

The second image is invisible because the value of each pixel is between 0 and 3. For any given pixel, its value is added to the to the collocated pixel in the first image that has a value from the set $\{0, 4, 8, \dots, 252\}$. The 2nd image is noise on the 1st.

Image 1 in upper 6-bits.
Image 2 in lower 2-bits.

Application of Quantization: Steganography

9-Shift-L

?

Pieter Bruegel (the Elder, ca. 1525-69), *The Peasant Dance*, 1568, Oil on oak panel, 114x164 cm, Kunsthistorisches Museum Wien, Vienna

To recover the second image (which is 2 bits per pixel per band) simply left shift the combined image by 6 bits.

Image 2 in upper 2-bits.
Image 1 shifted out

Application of Quantization: Steganography



From the video game, *Zero Wing*, by Toaplan.
See http://en.wikipedia.org/wiki/All_your_base

To recover the second image (which is 2 bits per pixel per band) simply left shift the combined image by 6 bits.

Remote Sensing: multiband Images (Landsat)



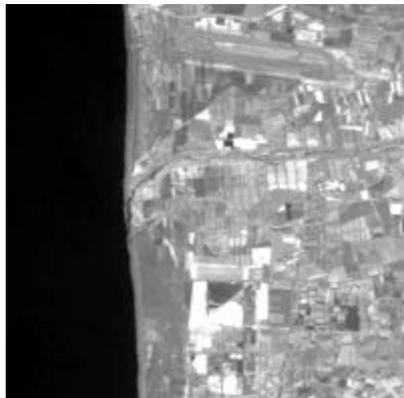
Band 1



Band 2



Band 3



Band 4



Band 5



Band 7

Remote Sensing: multiband Images

- We can visualize 3 bands at a time: pseudocolor

True Color



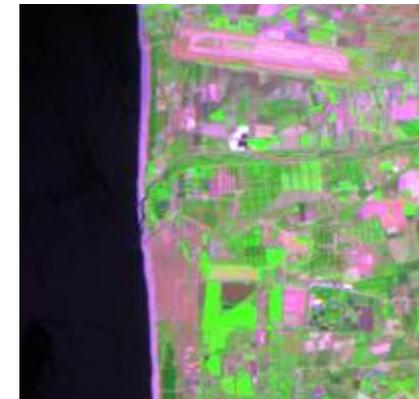
R G B
3 2 1

False Color Composite



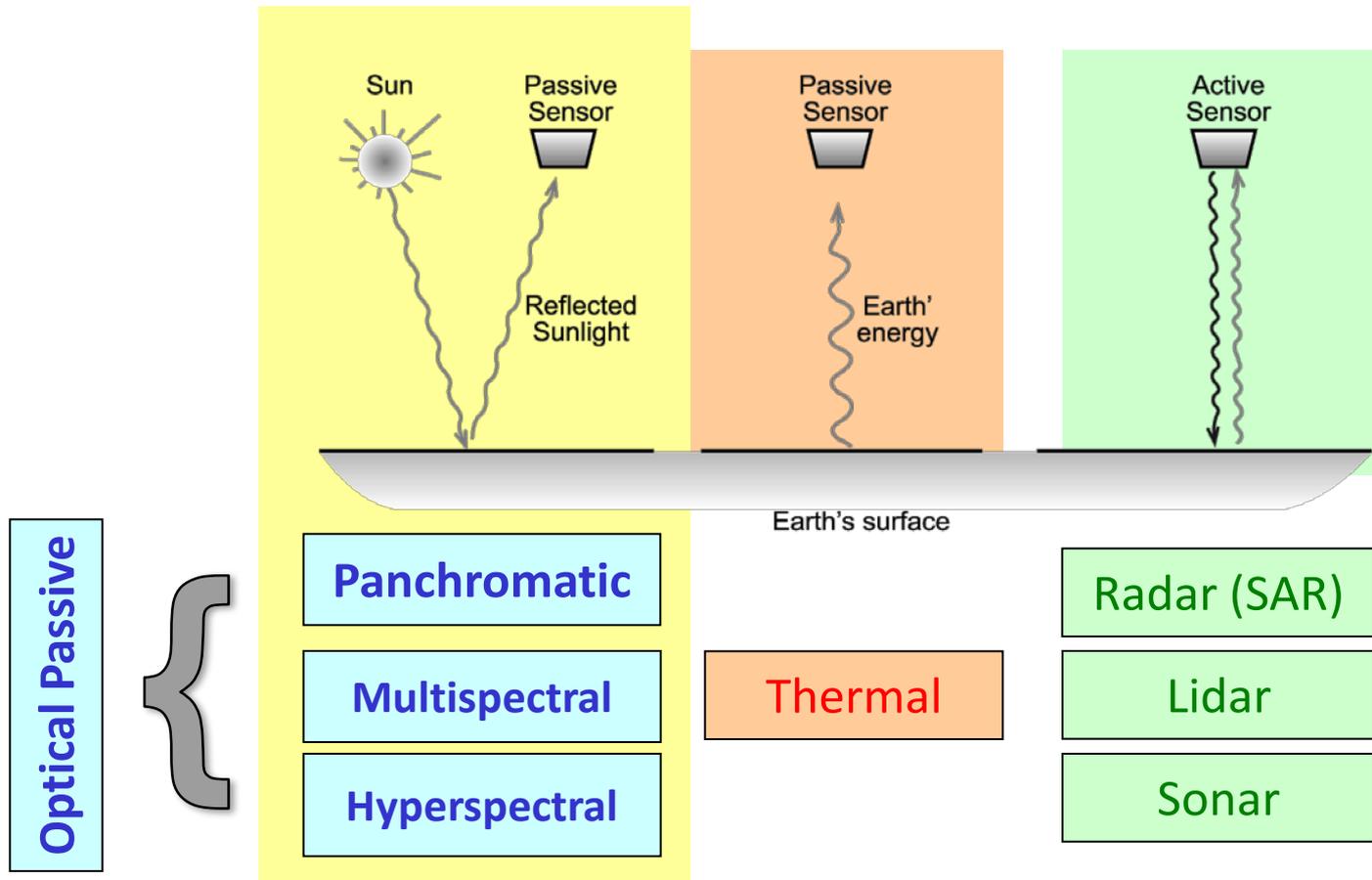
NIR R G
4 3 2

False Color Composite

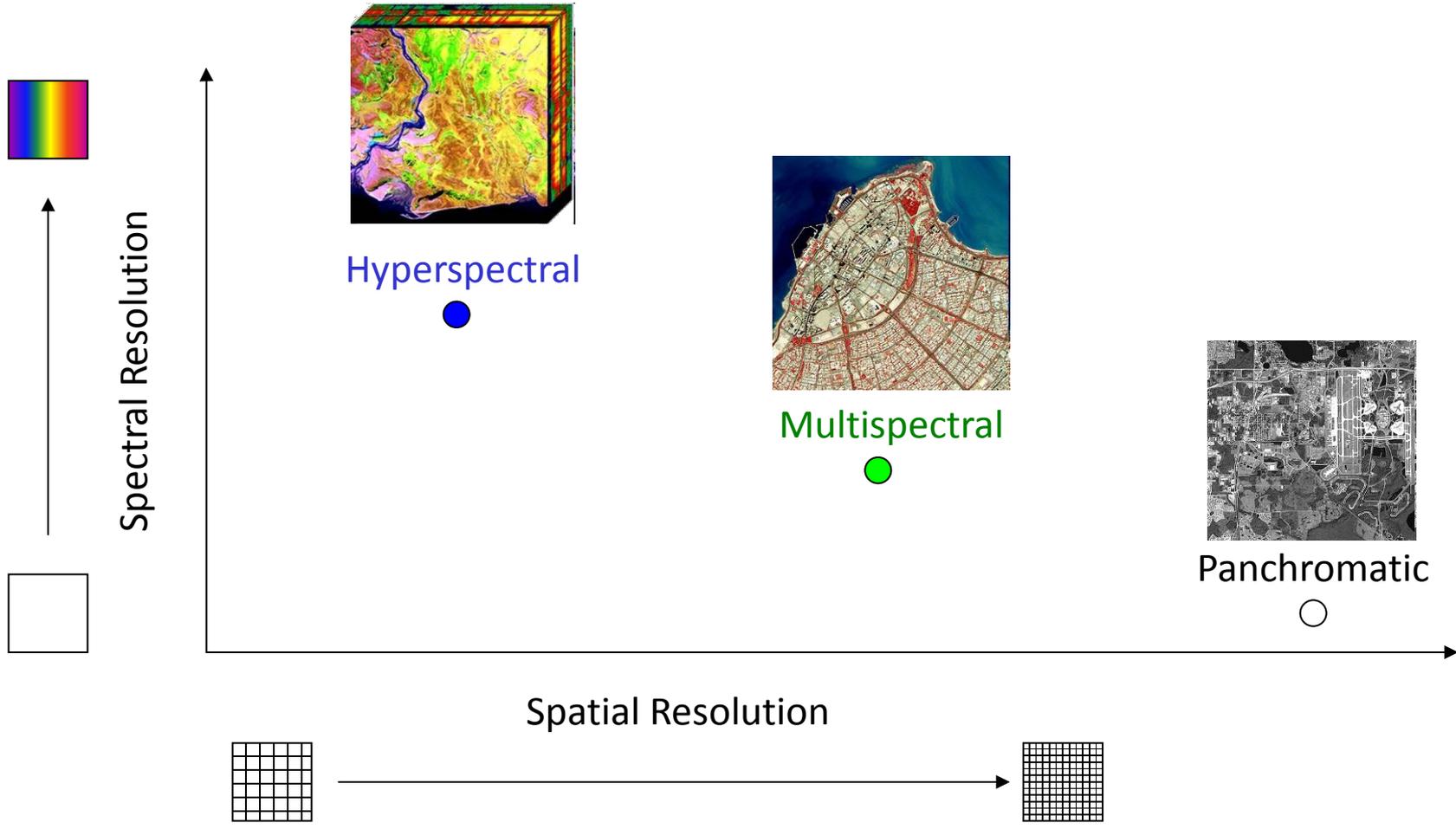


SWIR NIR R
7 4 3

How many Sensors / kinds of images / datasets in RS?



Optical Passive Sensors in Remote Sensing



Why are **Spatial** and *spectral*
resolution inversely proportional?

Hyperspectral



Buddingtonite



CHALCIDITE!



Alunite

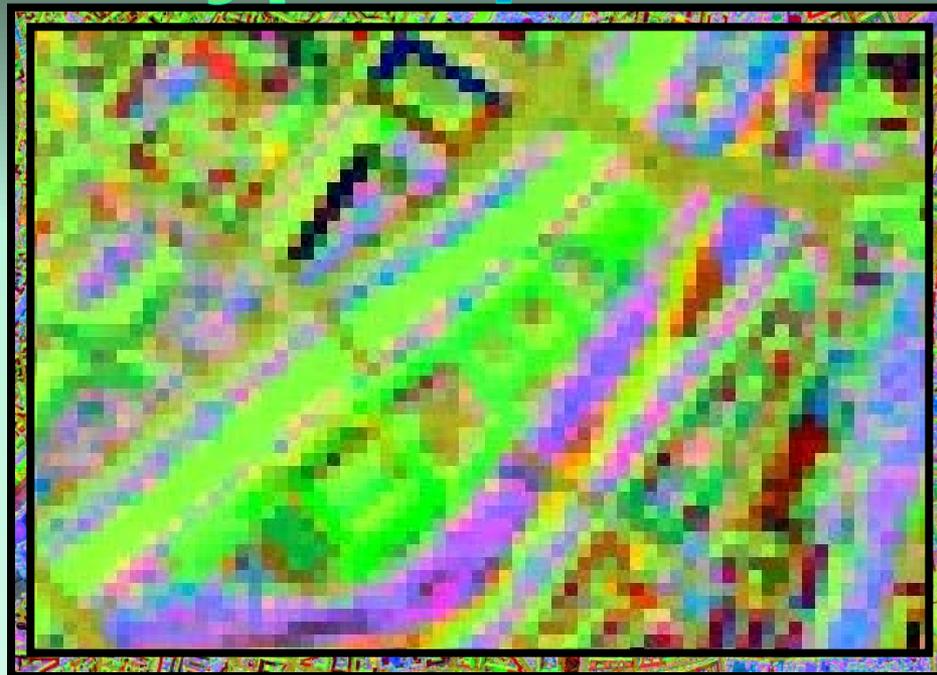


In the city..

Zoom in!

Panchromatic

Hyperspectral





Panchromatic

Hyperspectral



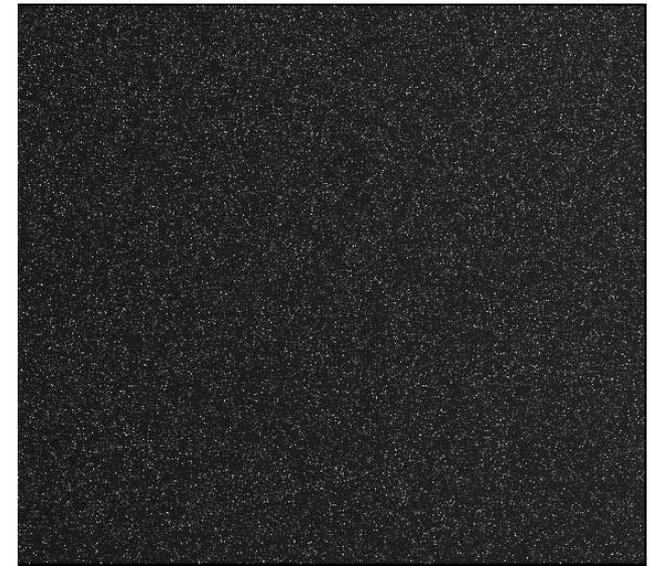
100 W

Image Correction

- Raw images are minimally processed images coming directly from the image sensors
- They usually go through several correction steps
- Some important ones are:
 - Dark Signal Correction
 - Non-linearity Correction
 - Odd-even Effect Correction
 - Dead Pixels Flagging

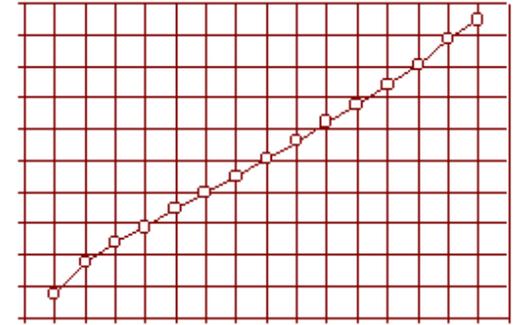
Dark Signal Correction

- Electronic interferences make the recorded signal (a bit) different than it really is
- Dark Signal Measurements
 - Shutter Method
 - Before every take an acquisition is made with the shutter closed. The resulting signal is the „dark current“
 - Deep Space Looking
 - Measures thermal radiation that can affect Dark Signal measurements
 - Dark pixels of the SWIR detector
 - Dark signal depends on the stability of the supplied voltage
 - During image special pixels which stay dark are used as reference



Dark Signal Measurement

Non-linearity Correction



- The response of a detector as a function of integration time is not linear
- The (non-linear) pixel response is measured at different exposition times and a correction is estimated as a linear function
- During the process dark signal has to be taken into account

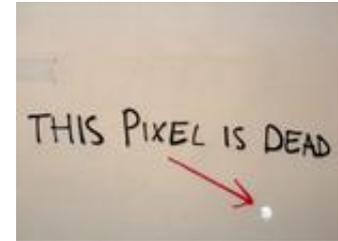
Odd-Even Effect

- The odd-even column effect consists of variations of the signal between the columns of the array
 - It is due to differences between sensor arrays
- It is easy to correct
 - Check the difference between the average values of a given column and its neighbouring columns



Raw Image with Odd-even Effect

Dead Pixel Map



- A list of pixels which readings do not have any meaning
- They are declared as „dead“ and ignored (set to 0)
- Different kinds of dead pixels:
 - No response
 - Very large output (hot pixel), saturates easily
 - Flickering pixel (constantly changing output)
 - Constant output

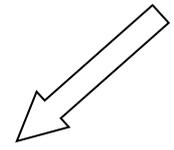


Image Correction

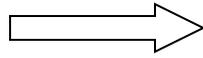
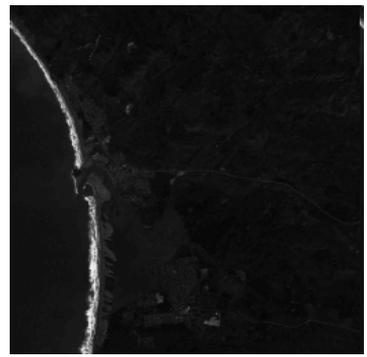
- Once our raw data are corrected, the image is formed and usually undergoes other correction steps....
 - Atmospheric Correction
 - Geometric Correction / Orthorectification..

- Image Acquisition & Correction
 - Raw Data → Raw Image → Image

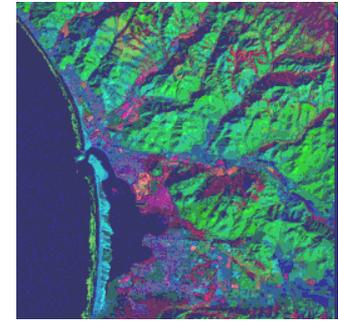
03	29	38	48
59	96	94	04
05	06	96	97
87	76	75	45



- Low-level Analysis
 - Image → Image
 - Time domain
 - Frequency domain



- Mid-level Analysis
 - Image → Features / Attributes
 - Feature Extraction
 - Clustering / Segmentation



- | | | | |
|--|---------------|--|---------------------------|
| | Beach Bar | | Urban Area |
| | Wave Breakers | | Shadows |
| | Vegetation1 | | Sea |
| | Vegetation2 | | Mountains (bright slopes) |
| | Golf Course | | |

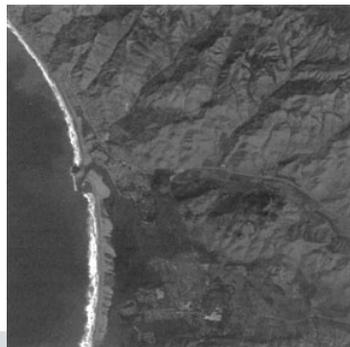
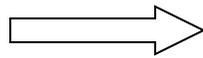
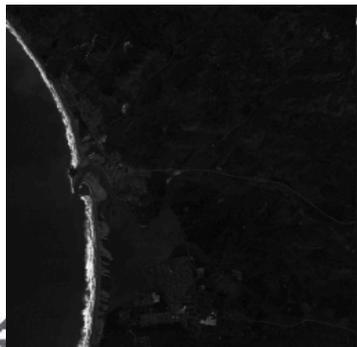
- High-level Analysis
 - Features → Recognition

Summary

- Image Acquisition & Characteristics
 - Spatial, radiometric & spectral resolution
 - Image Correction
- Image enhancement
 - Time Domain
 - Global Techniques: Histogram Stretch
 - Local Techniques: Moving Window Transform
 - Frequency Domain
- Sampling & Aliasing
- Image Features
- Image Clustering
- Image Classification

Image enhancement

- *Enhance*: to make greater (as in value, desirability, or attractiveness)
- The principal objective of enhancement is to process an image so that the result is more suitable than the original for a *specific* application
- Enhancement is subjective!
 - A good technique for a given application is not valid for another one

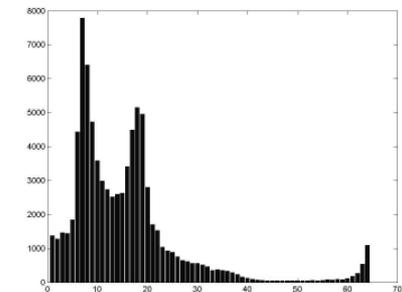
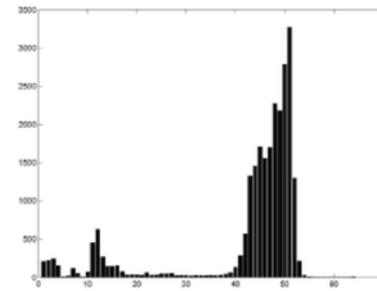
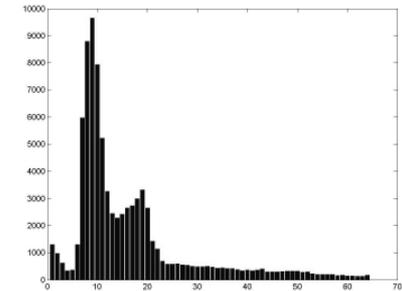
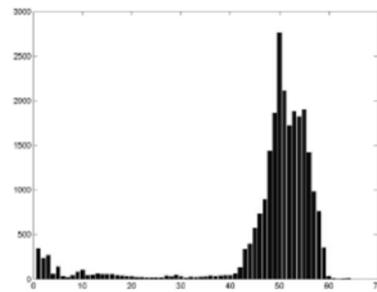


© Original Artist
 Reproduction rights obtainable from
www.CartoonStock.com

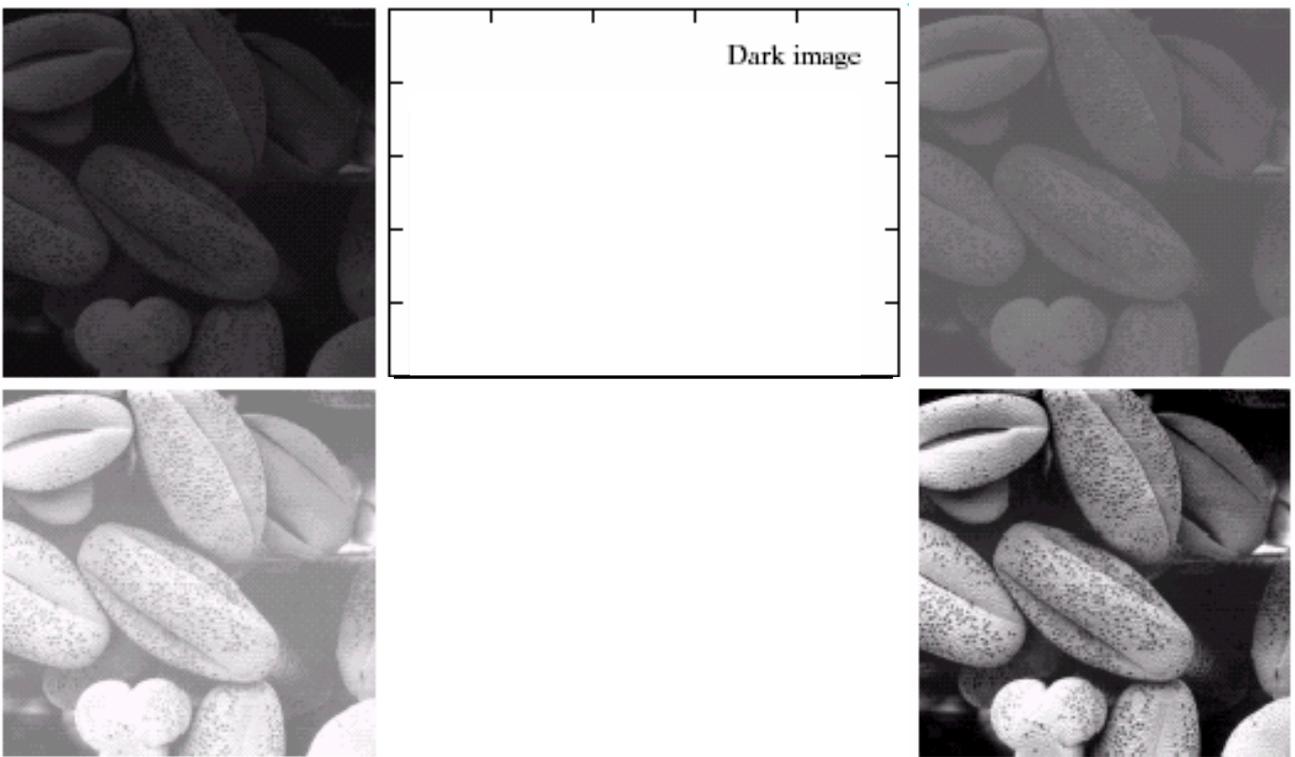


" These photo's are ruined..Not one of them has 'Red eye'!"

Sample Histograms, Natural Images



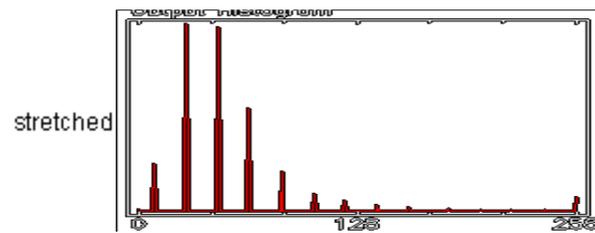
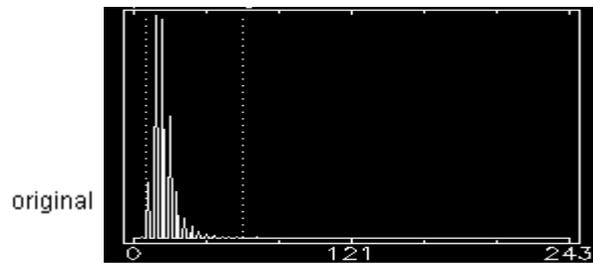
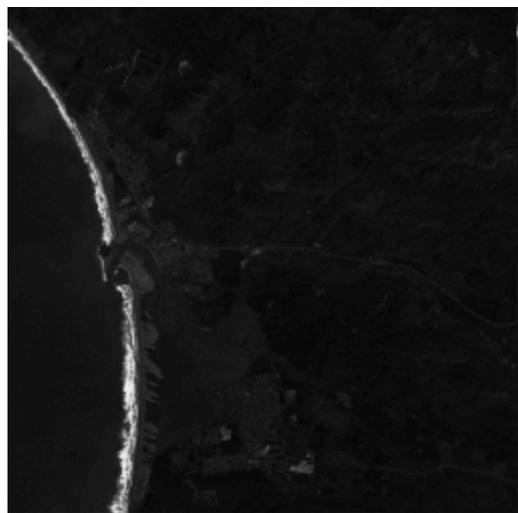
Histogram processing



How do you expect the histograms for these pictures?



Histogram Stretching



In how many ways can we stretch this?



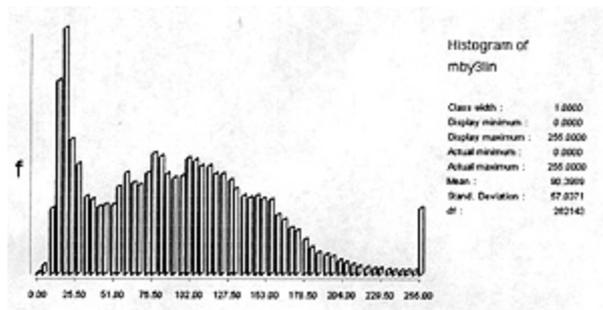
So?



Histogram Stretching

Selective Linear Stretch

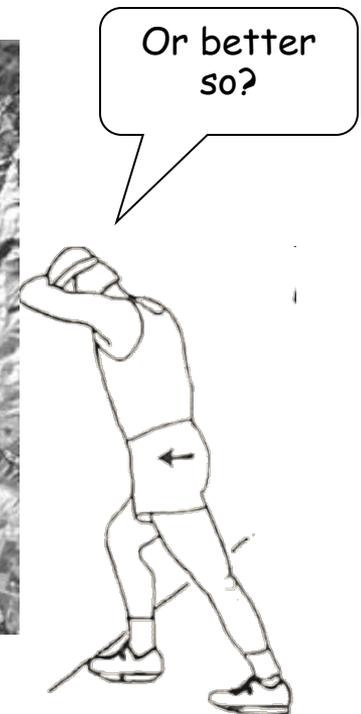
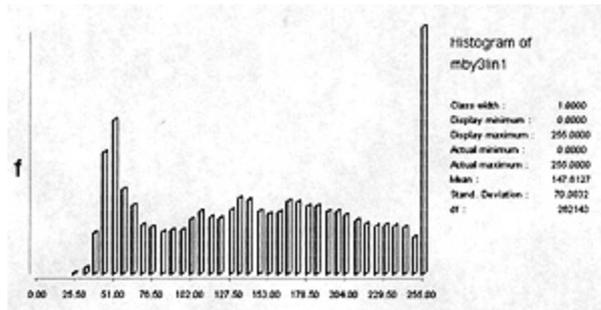
- We take Digital Numbers between 5 and 65
- We expand these from 0 to 255
- All values < 5 are set to 0
- All values > 65 are set to 255
- All values in between are stretched proportionally



Histogram Stretching

Selective Linear Stretch, let us try to get rid of these dark areas!

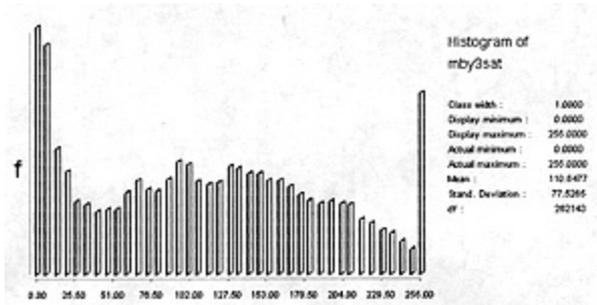
- We take now the DNs between 0 and 45
- We expand these from 0 to 255



Histogram Stretching

Linear-with-Saturation stretch

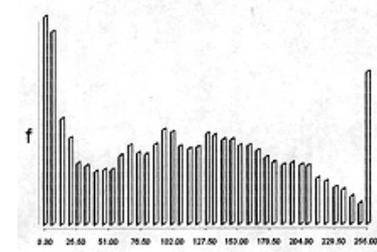
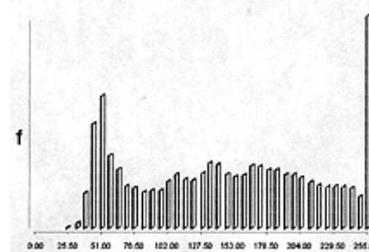
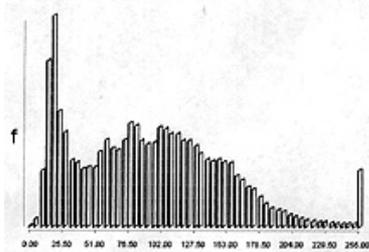
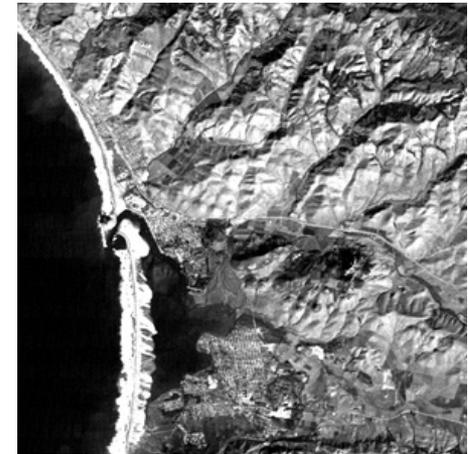
We assign 5% of pixels at each end (tail) of the histogram to single values, and stretch the values in between



That looks better...



Histogram Stretching: Comparison



Selective
Stretch I

Selective
Stretch II

Automatic
Linear-with-
Saturation



Linear Histogram Equalization



- All 4 images are mapped to a similar output image by applying the same histogram equalization function



Histogram Equalization vs. Linear Stretch

