Image Processing & Vision

Week 2





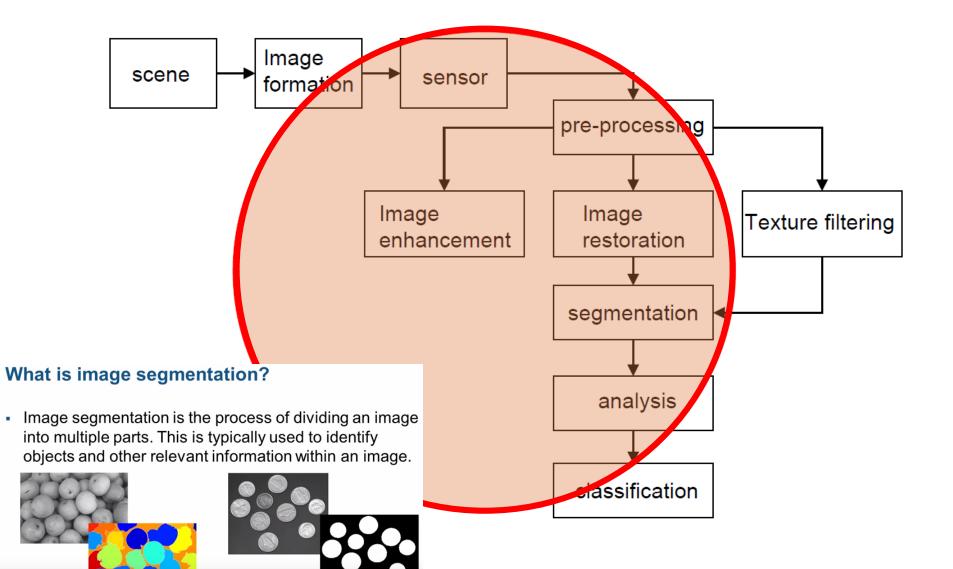
Content

- Recap:
 - Matlab, DIPimage, loading and display image

Image filtering 1

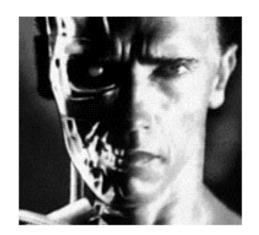


Image Analysis Paradigm



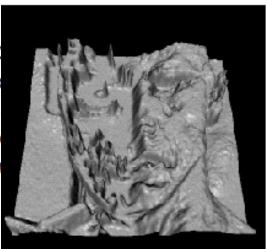
What are images...

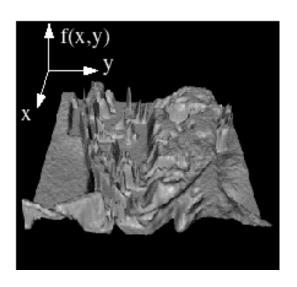
- An imag position
 - Radia
 - Acous
 - Surfa



of

- 2-D imc
 - Projec
- 3-D imo
 - CT: Co
 - MRI:
 - Optic
 - Seism

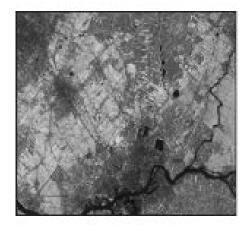




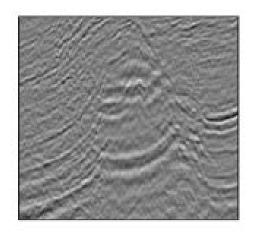


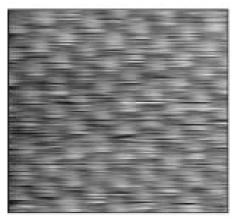
Examples of Images





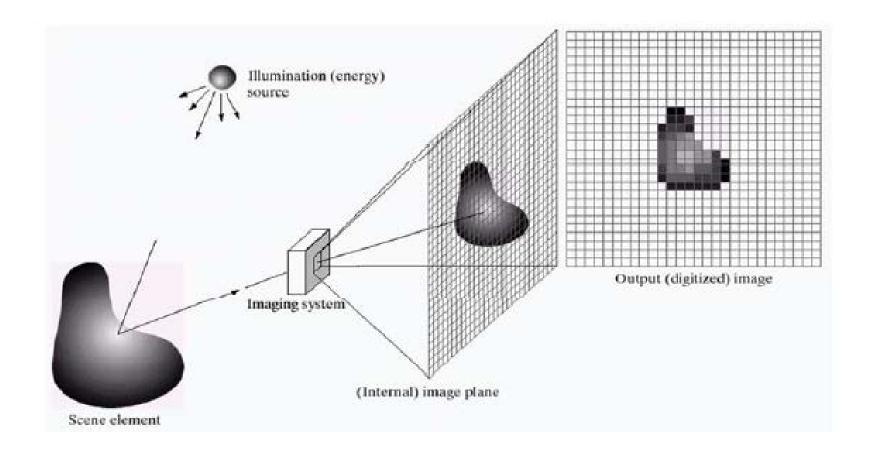








Digital Image Acquisition Process





Sampling and Quantization

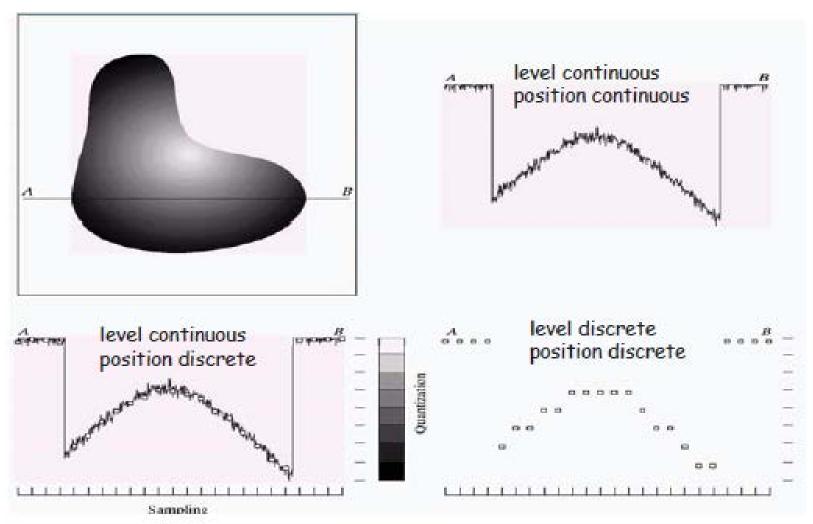
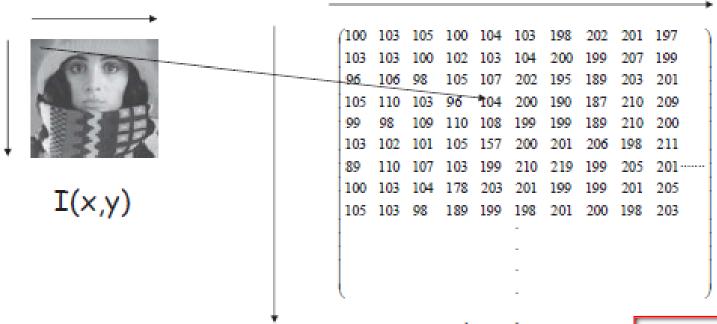


Image Representation

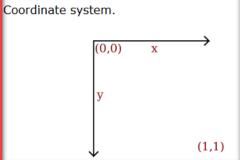
Image can be represented as a matrix



M(m,n)

$$I(5,4)=M(5,4)=104$$





Sampling



256x256



128x128



64x64

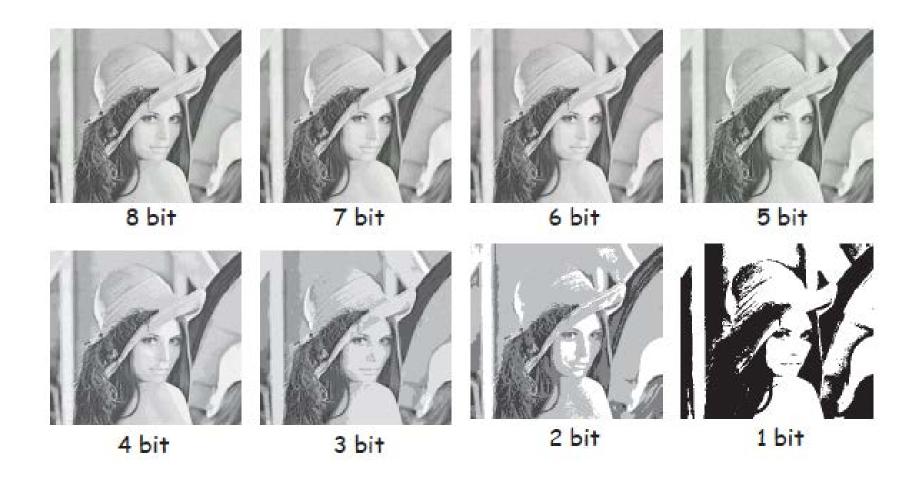


16x16

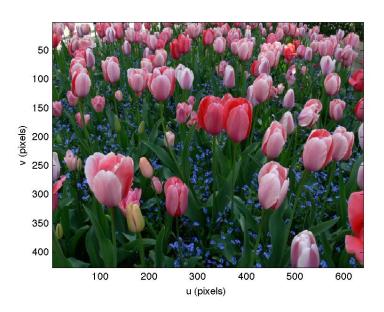
B



Effects of Quantization

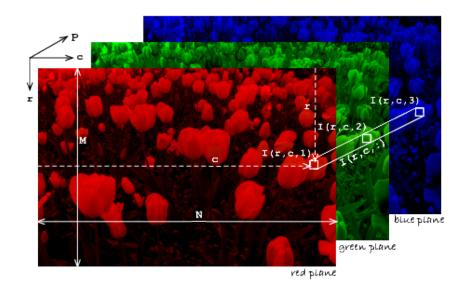


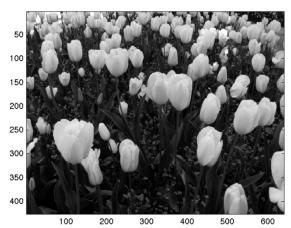
Color channels



multi-channel images (Blue, Green, Red)

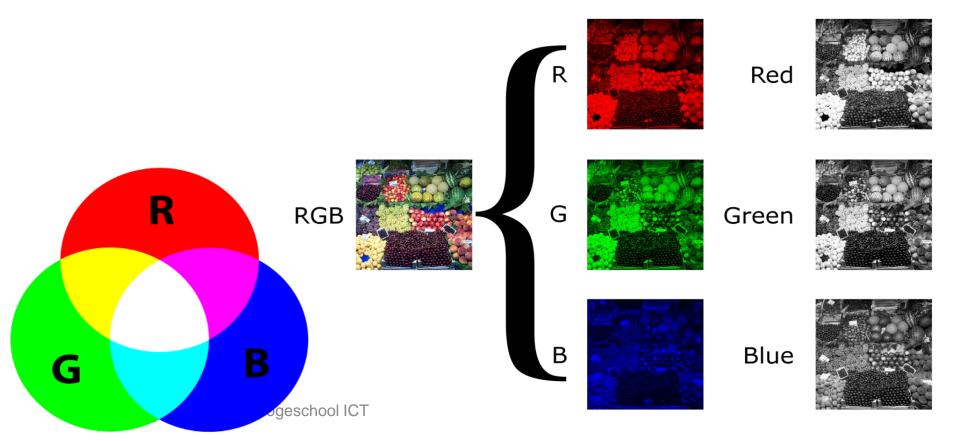






Example:

- color channel splitting of a full RGB color image
 - Here is The column at left shows the isolated color channels in natural colors, while at right there are their grayscale equivalence
 - https://nl.mathworks.com/help/matlab/creating_plots/image-types.html



Bit depth/channels

- Bit depth is the number of bits per pixel
- Quantifies how many unique colors are available in an image's color palette

Bits Per Pixel	Number of Colors Available	Common Name(s)
1	2	Monochrome
2	4	CGA
4	16	EGA
8	256	VGA
16	65536	XGA, High Color
24	16777216	SVGA, True Color
32	16777216 + Transparency	
48	281 Trillion	

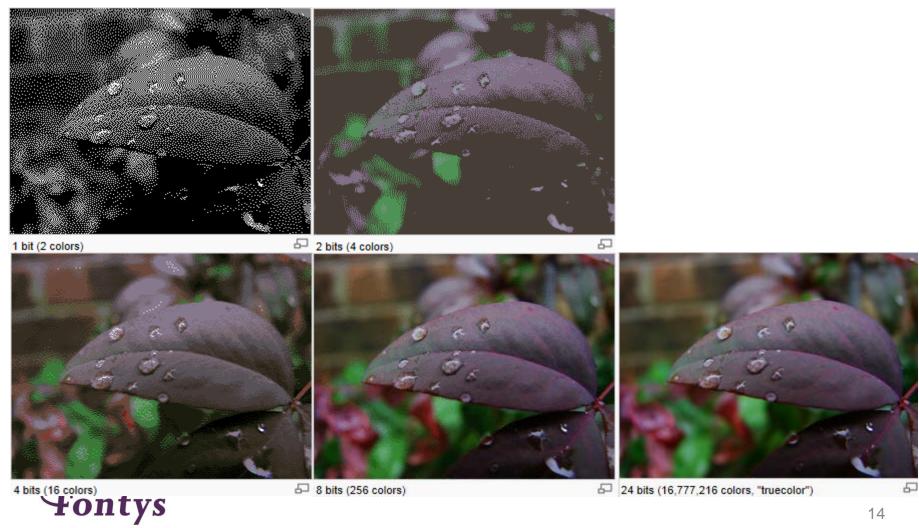
Every color pixel in a digital image is created through some combination of the three primary colors: red, green, and blue.



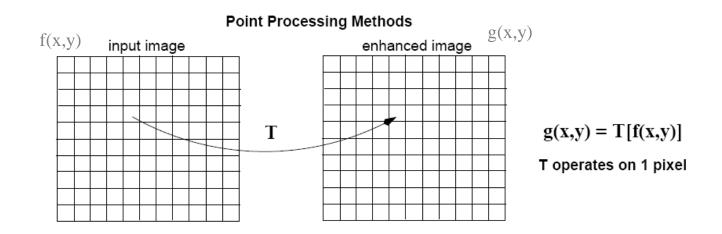
8 bits



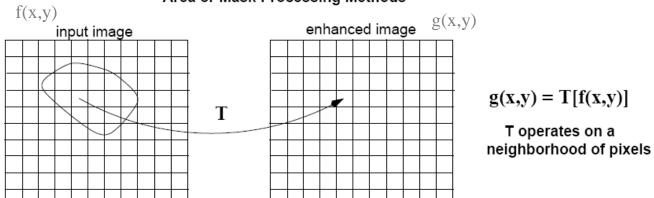
Bit depth



Spatial Domain Methods



Area or Mask Processing Methods

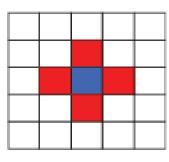




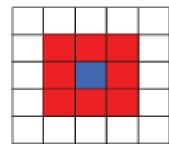
Connectivity



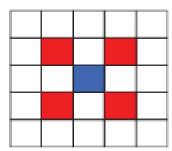
4-connected neighbours $N_4(p)$



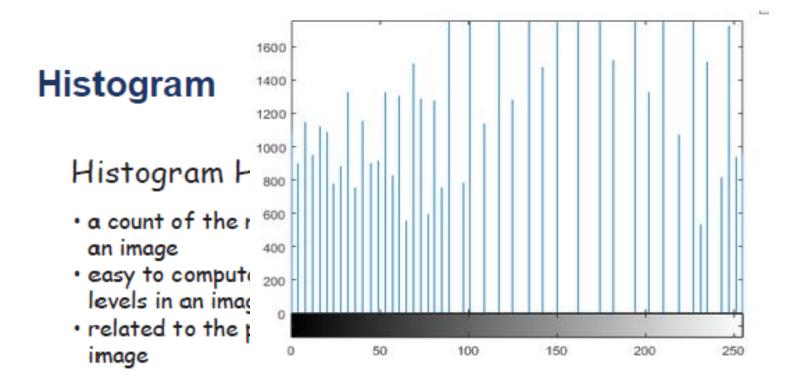
8-connected neighbours $N_8(p)$



Diagonal neighbours ND(p)





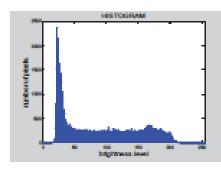


- H(i)=number of pixels in image with gray level I
- · Gray level sometimes called brightness or intensity
- Number of levels (also called number of bins) depends on gray level quantization (e.g. 8 bit image has 256 levels)
- H(i) is a discrete function
- H(i) ≥ 0 and H(i) ≤ #pixels in image for all i.

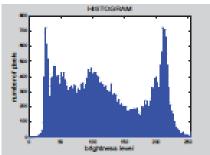


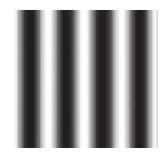
Histogram Examples

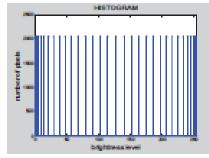






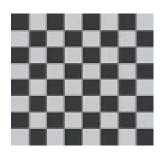


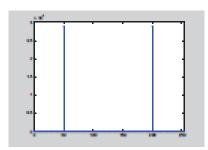


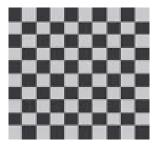


Histogram Properties

Every image has a unique histogram but not vice versa







Total number of pixels is sum over all brightness levels i

#pixels =
$$\sum_{i=0}^{i=max} H(i)$$





Image Filtering



Image Filtering

- Basic (low level) image processing operations:
 - Filtering
 - Enhancement
- Usually important pre-processing steps in computer vision systems (often in hardware)







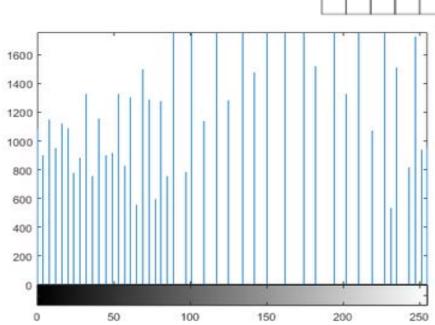
Content

- Linear and nonlinear spatial filtering
 - linear filters: smoothen/sharpen the edges in an image
 - Understand how the two operations are related
- Edge enhancement ← → edge detection

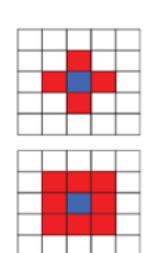


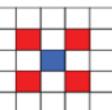
Image Filtering operations

- Simple image operations:
 - 'neighbourhood'
 - More general filtering operations use!!!
 - 'pointwise'
 - Histogram equalisation









Linear & non-linear filters

Basically there are two types of filters:

Linear Filters

- GaussianFilter
- DerivativeFilter
- MeanFilter
- GradientFilter ...

Nonlinear Filters

- MedianFilter
- MinFilter
- MaxFilter
- KuwaharaFilter



Linear filters can be implemented by a convolution

But non-linear filters can not.

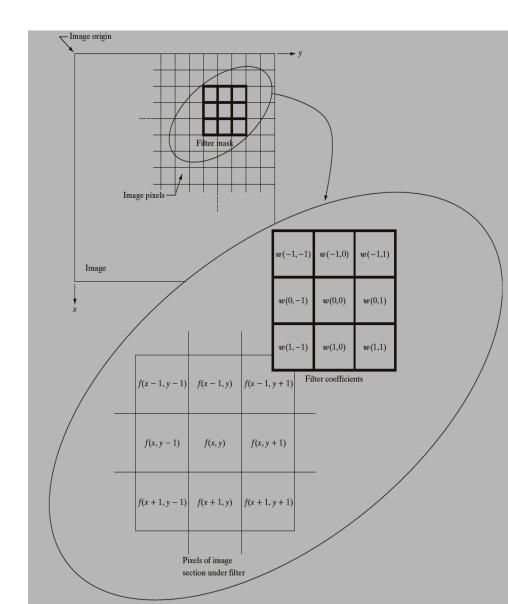
Convolution Kernels

- DiskMatrix
- BoxMatrix ...

Linear Operation

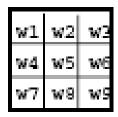
- Typically linear combinations of pixel values.
 - e.g., weight pixel values and add them together.
- Applications: by using different weights.
 - e.g., smoothing, sharpening, edge detection.

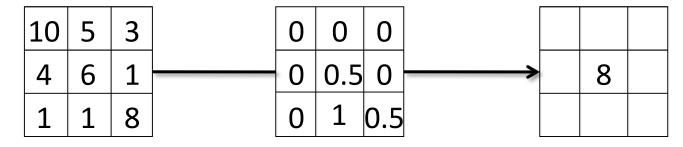






Example





Local image neighborhood

mask

Modified image data



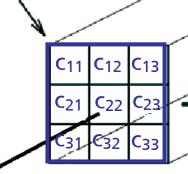
Convolution

$$R_{ij} = \sum_{u,v} H_{i-u,j-v} F_{uv}$$

'overlap-multiply -add' with 'convolution mask'

image f

convolution mask



$$o(i,j) = c_{11} f(i-1,j-1) + c_{12} f(i-1,j) + c_{13} f(i-1,j+1) + c_{21} f(i,j-1) + c_{22} f(i,j) + c_{23} f(i,j+1) + c_{31} f(i+1,j-1) + c_{32} f(i+1,j) + c_{33} f(i+1,j+1)$$

f(.)

f(.) |f(.) | f(.)

f(.) f(.)



i-1

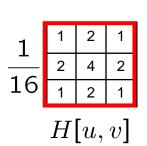
i

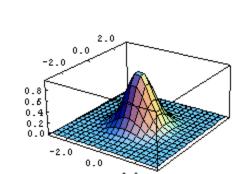
i+1

Linear filter: Gaussian Filtering(demo)

- Properties
 - Output is <u>a linear function</u> of the input
 - Output is <u>a shift-invariant function</u> of the input
- Gaussian kernel gives less weight to pixels further from the center of the window
- This kernel is an approximation of a Gaussian function:

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0





 $h(u,v) = \frac{1}{2\pi\sigma^2} e^{-\frac{u^2 + v^2}{\sigma^2}}$



F[x, y]

Demo

http://setosa.io/ev/image-kernels/





