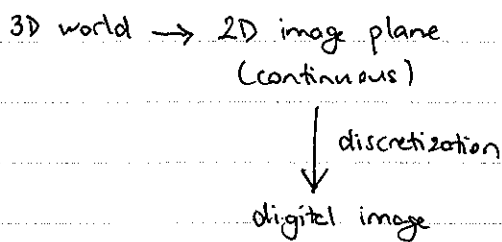
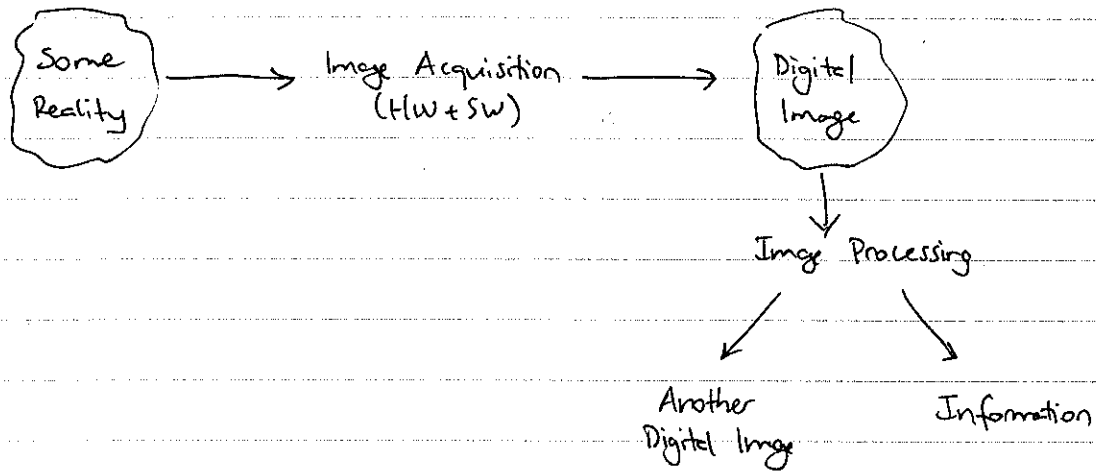


# Image Processing Basics



## Discretization in

1. space (sampling)
  2. brightness (quantization)
- $\rangle$  resolution  $\rightarrow$  picture quality

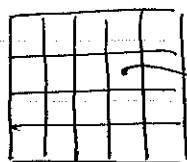
digital image: 2D array containing values from a finite set

## Examples:

- Bitmaps: boolean per pixel;  $I: \mathbb{R}^2 \rightarrow \{0, 1\}$  (black and white)
- Grayscale: integer per pixel; (typical)  $I: \mathbb{R}^2 \rightarrow [0, 255]$
- Color: 3 integers per pixel;  $I: \mathbb{R}^2 \rightarrow [0, 255]^3$
- Floating point: floating point per pixel;  $I: \mathbb{R}^2 \rightarrow \mathbb{R}_+$   
(3 floating points)  $I: \mathbb{R}^2 \rightarrow \mathbb{R}_+^3$

(HDR: high dynamic range images)

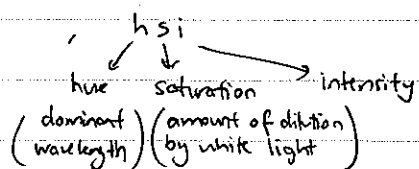
## Color images



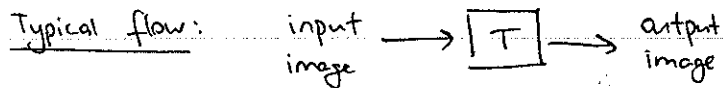
$(c_1, c_2, c_3)$

e.g.  $(r, g, b)$

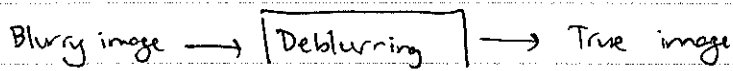
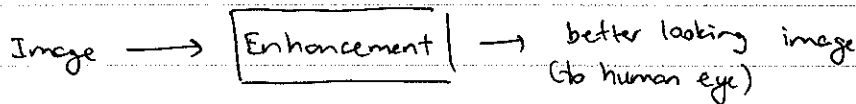
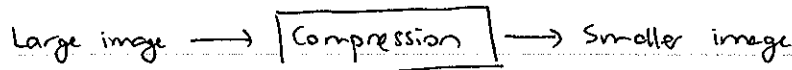
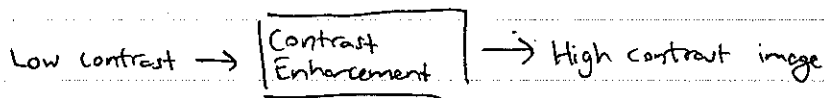
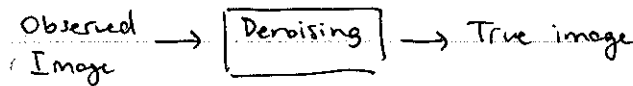
$\sim 700\text{nm}$     $\sim 500\text{nm}$     $\sim 400\text{nm}$



# Image Processing Basics



input image = true image + noise (adaptive noise model)  
(observed)

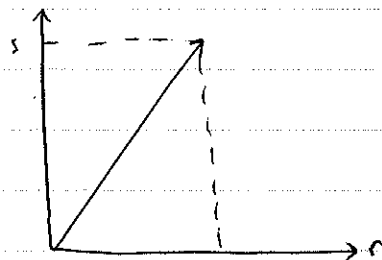


## Point operations

operations on single pixels

$$g(x) = T[f(x)]$$

output image      transformation function      input image



$$s = T(r)$$

r: previous value

s: new value

e.g. Contrast stretching

input image  $\theta$  in  $[0, 5]$

output image  $\theta$  in  $[0, 50]$

(Dynamic range)

# Image Processing Basics

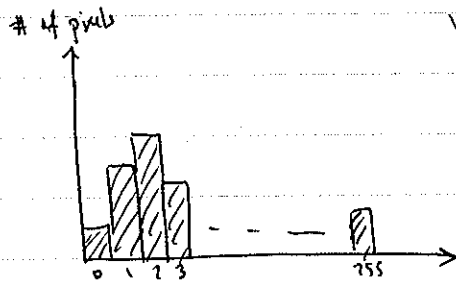
Histogram Equalization: A good quality image has a nearly uniform distribution of intensity levels

histogram: brightness distribution

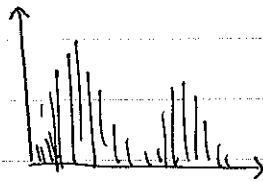
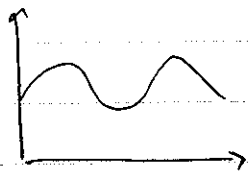
- a discrete function  $h(r_k) = n_k$

$\downarrow$   
 $k^{\text{th}}$  intensity value

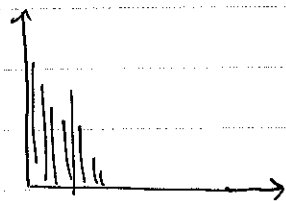
$\swarrow$   
# of pixels in the image with intensity  $r_k$



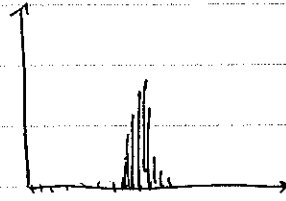
What histograms say about images?



object and background



dark



low contrast

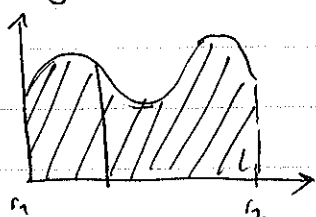
What they don't?

- no spatial information

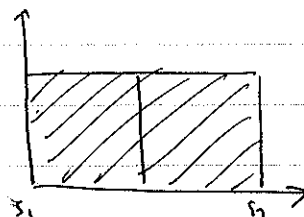


-----

Histogram Equalization



→



All values are equally likely

# Image Processing Basics

## Visual Enhancement

1. Contrast :- a property based on human perception abilities
  - local change in brightness
  - under good conditions, we can notice a 2% change in intensity (Gamma correction in display devices)

2. Noise: True signal + noise = observation (Additive noise model)

- Gaussian noise:  $N \sim N(0, \sigma)$

- Impulse noise

  - eg. salt/pepper noise

- !

• In Matlab,

- `randn`: generates a random variable from a normal distribution



- `rand`: generates a r.v. from a uniform distribution

• In additive noise model,

- there are two components:

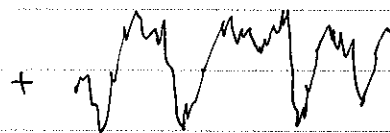


smooth/piecewise smooth

- low frequency components

↓ why smooth?

neighboring pixels  
have similar brightness



oscillating

- high frequency components

↓ why oscillating?

local changes  
in brightness

- noise

- image edges

# Image Processing Basics

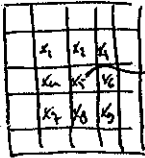
## Arithmetic mean filter

: averaging in a neighborhood

$$\hat{f}(x,y) = \frac{1}{mn} \sum_{(s,t) \in S_{xy}} g(s,t)$$

(low pass filtering)

↓  
neighborhood  
window



measurement at pixel  $i$  = true value at  $i$  + noise at  $i$

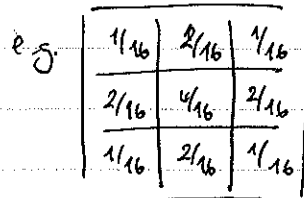
search for the true value  $c$

find the least-square estimate  $\hat{c}$  for  $c$

$$\min_{c} \sum_{i=1}^N (x_i - c)^2 \quad N=9$$

$$\Rightarrow c = \frac{1}{N} \sum_{i=1}^N x_i \quad (\text{mean})$$

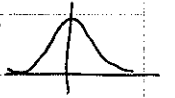
weighted averaging:  $\hat{f}(x,y) = \frac{1}{mn} \sum_{(s,t) \in S_{xy}} w(s,t) \cdot g(s,t)$



weights decay as a function of distance from the pixel of interest

why?

e.g. select the weights from a Gaussian  $\frac{1}{2\pi\sigma} e^{-\frac{1}{2\pi}(x-\mu)^2}$



Median filter  $\hat{f}(x,y) = \text{median} \{g(s,t)\}$   
 $(s,t) \in S_{xy}$

$$\Rightarrow \min \sum_{i=1}^N |x_i - c|$$

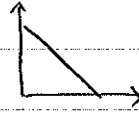
(robust to outliers  $\rightarrow$  salt/pepper noise)

# Image Processing Basics

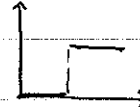
## Processing Techniques

### 1. Point Operations

a) Pixel based ex1: image negatives



ex2: thresholding



ex3: mid-tone enhancement

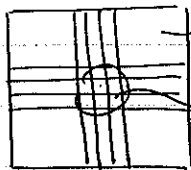
ex4:  $s = \log(r)$  (dynamic range enhancement)

ex5: Gamma-correction

b) Global techniques ex1: histogram equalization

ex2: histogram matching

### 2. Local techniques



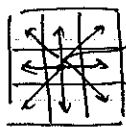
no relation

related to neighboring values

1. fixed neighborhood

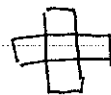
2. adaptive neighborhood

### Neighborhoods



$p(x,y)$

$N_4(p)$ : 4-neighborhood



$N_D(p)$ : diagonal



$N_8(p)$ :  $N_4(p) + N_D(p)$ : 8-neighborhood

### Distance measures

Let  $p, q$ , and  $z$  denote pixels with coordinates  $(x,y)$ ,  $(s,t)$  and  $(v,w)$ , respectively

$D$  is a distance function or metric if

i.  $D(p,q) \geq 0$  ( $D(p,q) = 0$  iff  $p=q$ ) (positivity)

ii.  $D(p,q) = D(q,p)$  (symmetry)

iii.  $D(p,z) \leq D(p,q) + D(q,z)$  (triangle inequality)

# Image Processing Basics

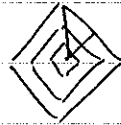
ex1. Euclidean Distance ( $L_2$ )

$$D_{L_2}(p, q) = \sqrt{(x-s)^2 + (y-t)^2}$$



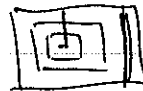
ex2. Manhattan Distance ( $L_1$ )

$$D_{L_1}(p, q) = |x-s| + |y-t|$$



ex3. Chessboard Distance ( $L_\infty$ )

$$D_{L_\infty}(p, q) = \max(|x-s|, |y-t|)$$



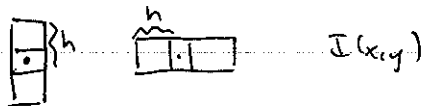
Edges: abrupt changes in the intensity

- ↳ edges (some structure)
- ↳ noise (isolated)

How to compute?

- use differentiation

- discrete derivatives



$$h=1$$

$$\frac{\partial I(x,y)}{\partial x} \Big|_{ij} = \frac{I(i+h,j) - I(i-h,j)}{2h}$$

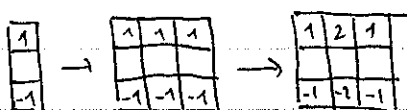
- central difference



- Forward difference



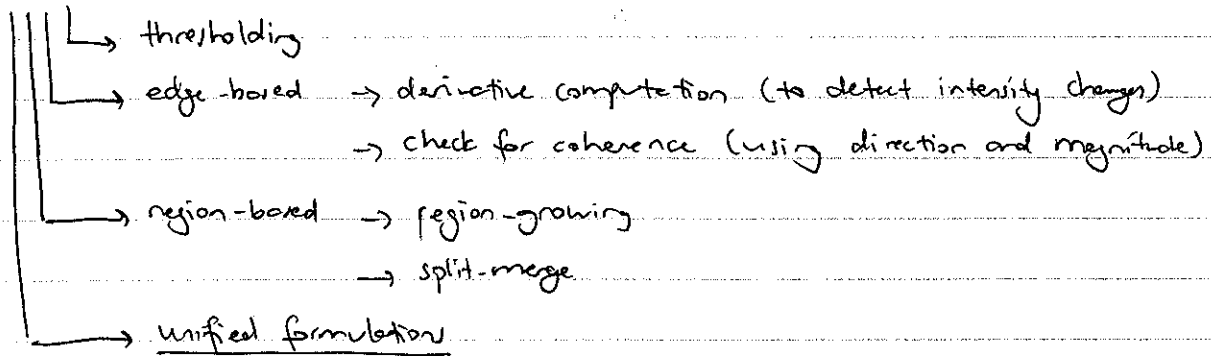
- Backward difference



Sobel operator

# Image Processing Basics

Segmentation: partitioning an image into coherent & regions



## Today's outline

- image representation
- point operations
- histogram processing
- contrast
- intensity variation (derivative computation)
- noise
- edge / boundary
- low/high frequency
- low-pass filtering
- averaging / median
- neighborhood
- distance