

Tutorial 2

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COMP435p
Biometrics Authentication

January 30, 2012





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- Problem 1: Answer the questions
- Problem 2: Power law function
- Problem 3: Convolution
- Problem 4: Noise reduction
- Problem 5: Filter operations



Outline

1 Problems

- Problem 1: Answer the questions
- Problem 2: Power law function
- Problem 3: Convolution
- Problem 4: Noise reduction
- Problem 5: Filter operations



Problem 1.a

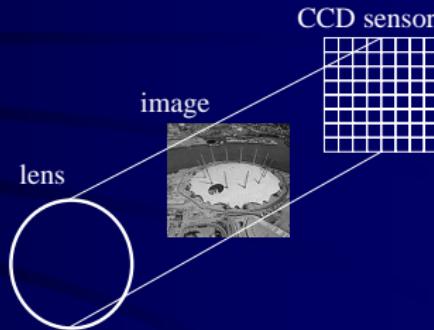
Understand some useful definitions in image processing: i) pixel, ii) image, iii) image resolution, d) gray-level image.



Problem 1.a

Digital Image Representation

- Optical image denoted by $A(x,y)$
- x and y are spatial co-ordinates
- $A(x,y)$ is the brightness at position (x,y)

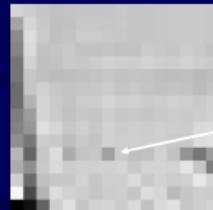


The points at which an image is sampled are known as *picture elements*, or *pixels*.



Problem 1.a

Digital Image Representation



picture
element
(pixel)

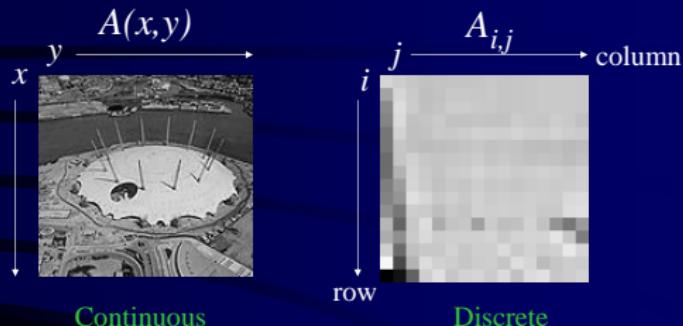
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Lecture 3- 7



Problem 1.a

Digital Image Representation



- An image is a spatial presentation of an object –
Matrix representing quantized intensity values.



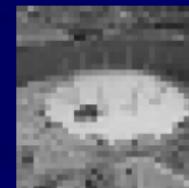
Problem 1.a

Image Resolution

- Image resolution: Measure of image quality (No. of pixels, $M \times N$)
 - 1280 x 960 (digital camera)
 - 512 x 512 (video camera)
 - 256 x 256 (reasonable processing quality)
- Dynamic range: Measure of the range of brightness values
 - 1 bit (binary image)
 - 8 bits (grey image: 0 for black, 255 for white)
 - 12 bits (medical imagery)
 - 16 bits (Astronomical imagery)
 - 24 bits (colour image)



High



Low



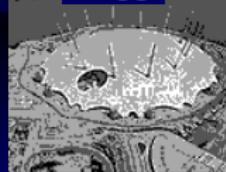
Problem 1.a

Gray-Level Quantisation

8 bpp



2 bpp



- For black and white images, a pixel is a single value integer or *floating point*.
- For display, normally 8 bits per pixel (bpp) is used.
- The human eye cannot resolve to this accuracy. 32 gray levels are usually sufficient (5 bpp).
- At 4 bpp and below, “false contouring” can become apparent.





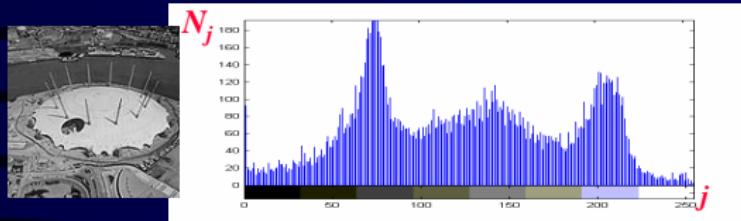
Problem 1.b

What is image histogram? What are peaks and plains in image histogram? Compare the differences between two types of bimodal histograms.



Problem 1.b

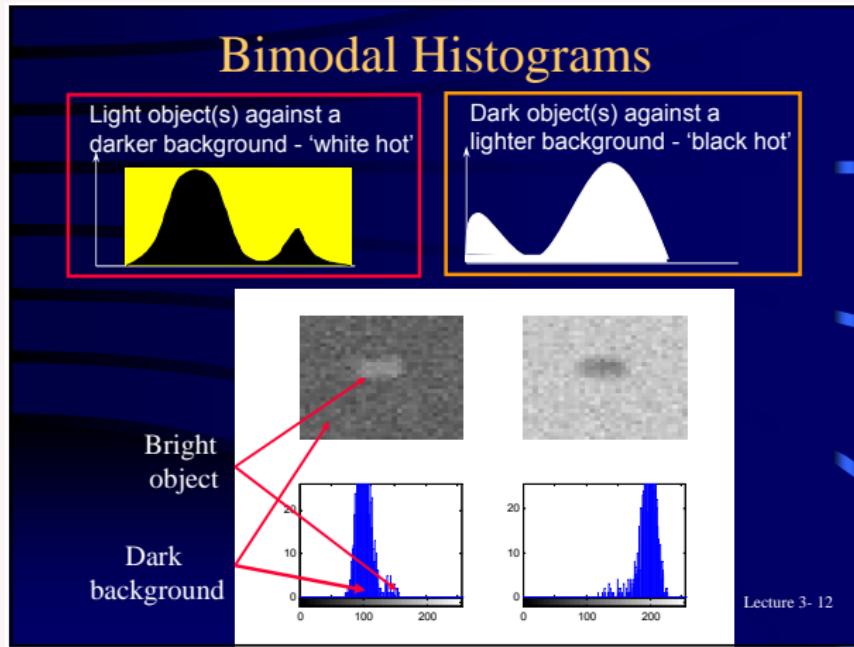
Image Histograms



- Plot of N_j versus j : Shows the distribution of image pixels in terms of their gray levels.
- Gray levels $j=0,1,\dots,255$
- N_j = number of pixels in image with gray level j



Problem 1.b





Problem 1.c

Understand the point operation. Contrast enhancement is the main application of pixel operation. How many methods of contrast enhancement do you learn? Please compare them.



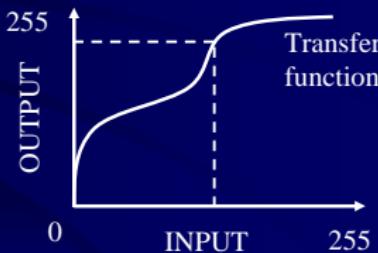
Problem 1.c

Point (Pixel) Operation

- Point operation: A function is applied to every pixel in an image, which operates only on the pixel's current value.
- Thresholding - A mask may be created by setting a pixel value to 1 or 0 depending upon if the current value is above or below a certain threshold value.

Input pixel value, I ,
mapped to output
pixel value, O , via
transfer function T .

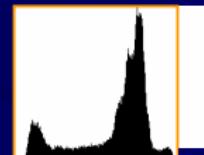
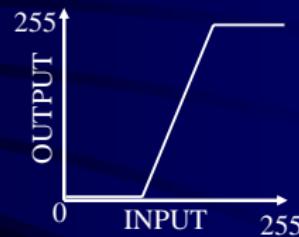
$$O = T(I)$$





Problem 1.c

Contrast Enhancement: Linear Stretching



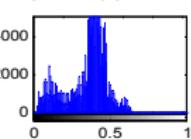
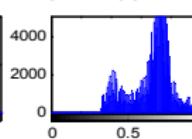
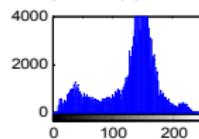
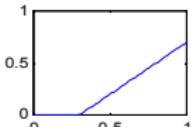
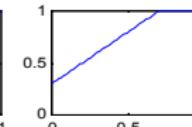
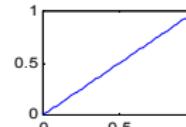
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Problem 1.c

Brightness



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Problem 1.c

Contrast Enhancement: Power Law Function

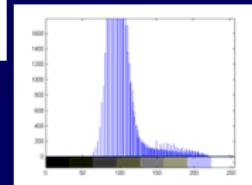
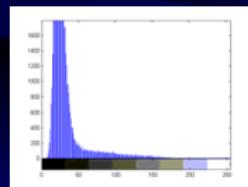
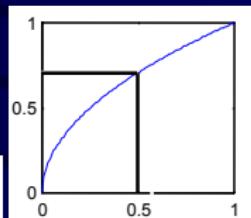
$$O = I^\gamma$$

- $\gamma < 1$ to enhance contrast in dark regions
- $\gamma > 1$ to enhance contrast in bright regions.



Problem 1.c

Contrast Enhancement: Power Law Function ($\gamma = 0.5$)



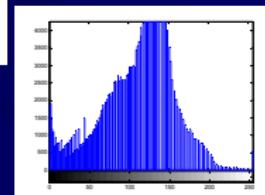
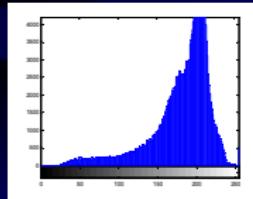
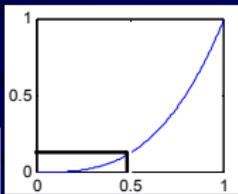
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Lecture 3- 18



Problem 1.c

Contrast Enhancement: Power Law Function ($\gamma = 3.0$)



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Lecture 3- 19



Problem 1.c

Contrast Enhancement: Power Law Function

- Look-up Table
 - Transfer function implemented as a look-up table (LUT).
 - Implemented in hardware or software.

$$\gamma=2$$

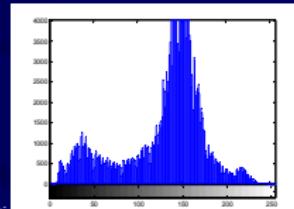
I	O
0	0
0.10	0.01
0.20	0.04
0.30	0.03
0.40	0.16
0.50	0.25
0.60	0.36
0.70	0.49
0.80	0.64
0.90	0.81
1.00	1.00



Problem 1.c

Contrast Enhancement: Histogram Equalisation

- Image histograms consisting of peaks and low plains.
- Peaks = many pixels concentrated in a few grey levels
- Plains = small number of pixels distributed over a wider range of grey levels



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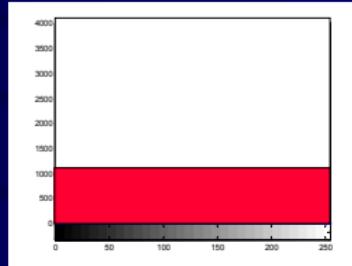
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Problem 1.c

Histogram Equalisation

- Expand pixels in peaks over a wider range of gray-levels.
- “Squeeze” low plains pixels into a narrower range of gray levels.
- Flat histogram.





Problem 1.c

Contrast Enhancement: Comparison

Original



$$\gamma > 1$$



Histogram
equalisation



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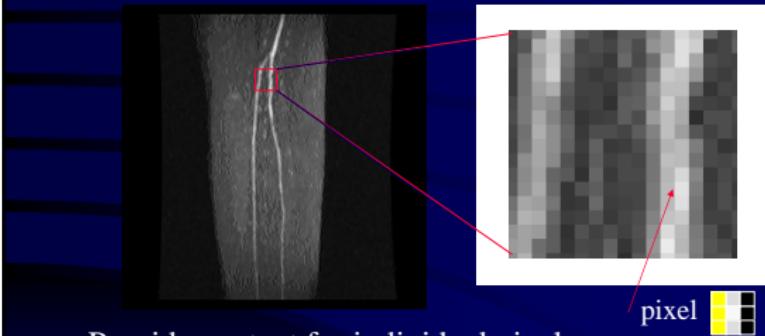
Problem 1.d

(d) Why neighbourhood operations? Give the applications of group operations.



Problem 1.d

Why are Neighbourhoods Important?



- Provide context for individual pixels.
- Relationships between neighbours determine image features.



Problem 1.d

Neighbourhood Operations

Noise Reduction



Edge Enhancement



Zooming



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Lecture 3- 26





Outline

1 Problems

- Problem 1: Answer the questions
- **Problem 2: Power law function**
- Problem 3: Convolution
- Problem 4: Noise reduction
- Problem 5: Filter operations



Problem 2: Power law function

What is the power law function? Give the LUT when $\gamma = 0.5$ in terms of the table in P3-20. Compare the two LUT and understand the impact of the parameter $\gamma = 0.5$.



Problem 2: Power law function terms

Contrast Enhancement: Power Law Function

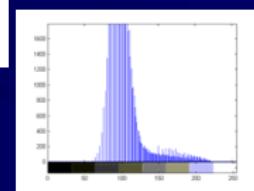
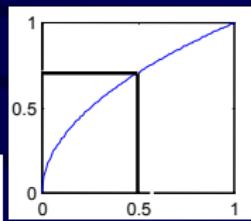
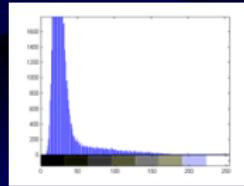
$$O = I^\gamma$$

- $\gamma < 1$ to enhance contrast in dark regions
- $\gamma > 1$ to enhance contrast in bright regions.



Problem 2: Power law function terms

Contrast Enhancement: Power Law Function ($\gamma = 0.5$)



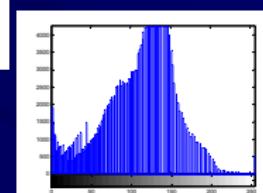
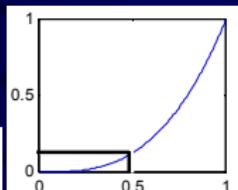
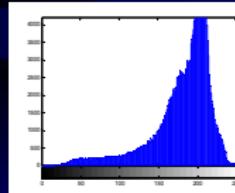
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Problem 2: Power law function terms

Contrast Enhancement: Power Law Function ($\gamma = 3.0$)



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Lecture 3- 19



Problem 2: Power law function terms

Contrast Enhancement: Power Law Function

$$\gamma=2$$

- Look-up Table
 - Transfer function implemented as a look-up table (LUT).
 - Implemented in hardware or software.

I	O
0	0
0.10	0.01
0.20	0.04
0.30	0.03
0.40	0.16
0.50	0.25
0.60	0.36
0.70	0.49
0.80	0.64
0.90	0.81
1.00	1.00



Problem 2: Power law function

$$I = \begin{bmatrix} 0 & 0.10 & 0.20 & 0.30 & 0.40 & 0.50 \\ 0.60 & 0.70 & 0.80 & 0.90 & 1.00 & \end{bmatrix}$$

$$O = I^\gamma$$

$$\begin{aligned} O &= \begin{bmatrix} 0^{0.5} & 0.10^{0.5} & 0.20^{0.5} & 0.30^{0.5} & 0.40^{0.5} & 0.50^{0.5} \\ 0.60^{0.5} & 0.70^{0.5} & 0.80^{0.5} & 0.90^{0.5} & 1.00^{0.5} & \end{bmatrix} \\ &= \begin{bmatrix} 0 & 0.32 & 0.45 & 0.55 & 0.63 & 0.71 \\ 0.77 & 0.84 & 0.89 & 0.95 & 1.00 & \end{bmatrix} \end{aligned}$$

□



Outline

1 Problems

- Problem 1: Answer the questions
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Problem 3: Convolution

Please check the convolution result in P3-33 and try to answer the question in P3-34. Now there is an input image and a mask as below. Could you give the convolved image?



Problem 3: Convolution terms

Convolution

- Consists of filtering an image A using a filter (mask) B .
- Mask is a small image whose pixel values are called weights.
- Weights modify relationships between pixels.

Filter,
mask or
template **B**

B _{1,1}	B _{1,2}
B _{2,1}	B _{2,2}

2×2

Input
image **A**

A _{1,1}	A _{1,2}	A _{1,3}	A _{1,4}
A _{2,1}	A _{2,2}	A _{2,3}	A _{2,4}
A _{3,1}	A _{3,2}	A _{3,3}	A _{3,4}
A _{4,1}	A _{4,2}	A _{4,3}	A _{4,4}

4×4

Convolved
Image **C**

C _{1,1}	C _{1,2}	C _{1,3}
C _{2,1}	C _{2,2}	C _{2,3}
C _{3,1}	C _{3,2}	C _{3,3}

3×3

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Problem 3: Convolution terms

Convolution

$B_{1,1}$	$B_{1,2}$
$B_{2,1}$	$B_{2,2}$
$A_{1,3}$	$A_{1,4}$
$A_{2,3}$	$A_{2,4}$
$A_{3,1}$	$A_{3,2}$
$A_{3,3}$	$A_{3,4}$
$A_{4,1}$	$A_{4,2}$
$A_{4,3}$	$A_{4,4}$

$A_{1,1} \times B_{1,1}$	$A_{1,2} \times B_{1,2}$
$A_{2,1} \times B_{2,1}$	$A_{2,2} \times B_{2,2}$

$$C_{1,1} = [A_{1,1} \times B_{1,1}] + [A_{1,2} \times B_{1,2}] + [A_{2,1} \times B_{2,1}] + [A_{2,2} \times B_{2,2}]$$



Problem 3: Convolution terms

Convolution

A _{1,1}	A _{1,2}	B _{1,B}	B _{1,A}
A _{2,1}	A _{2,2}	B _{2,B}	B _{2,A}
A _{3,1}	A _{3,2}	A _{3,3}	A _{3,4}
A _{4,1}	A _{4,2}	A _{4,3}	A _{4,4}

A _{1,3} × B _{1,1}	A _{1,4} × B _{1,2}
A _{2,3} × B _{2,1}	A _{2,4} × B _{2,2}

$$C_{1,3} = \boxed{A_{1,3} \times B_{1,1}} + \boxed{A_{1,4} \times B_{1,2}} + \boxed{A_{2,3} \times B_{2,1}} + \boxed{A_{2,4} \times B_{2,2}}$$



Problem 3: Convolution terms

Convolution

A _{1,1}	A _{1,2}	A _{1,3}	A _{1,4}
B _{1,1}	B _{1,2}	A _{2,3}	A _{2,4}
B _{2,1}	B _{2,2}	A _{3,3}	A _{3,4}
A _{4,1}	A _{4,2}	A _{4,3}	A _{4,4}

A _{2,1} × B _{1,1}	A _{2,2} × B _{1,2}
A _{3,1} × B _{2,1}	A _{3,2} × B _{2,2}

$$C_{2,1} = \boxed{A_{2,1} \times B_{1,1}} + \boxed{A_{2,2} \times B_{1,2}} + \boxed{A_{3,1} \times B_{2,1}} + \boxed{A_{3,2} \times B_{2,2}}$$



Problem 3: Convolution terms

Mathematical Notation

$$C_{1,1} = \boxed{A_{1,1} \times B_{1,1}} + \boxed{A_{1,2} \times B_{1,2}} + \boxed{A_{2,1} \times B_{2,1}} + \boxed{A_{2,2} \times B_{2,2}}$$

$$B = M \times N$$

$$C_{k,l} = \sum_{i=k}^{k+M-1} \sum_{j=l}^{l+N-1} A_{i,j} \times B_{i-k+1, j-l+1}$$

$$\sum_{i=1}^2 i = 1 + 2$$

$$\sum_{i=1}^2 A_i = A_1 + A_2$$

$$\sum_{i=1}^2 \sum_{j=1}^2 A_{i,j} = \sum_{i=1}^2 (A_{i,1} + A_{i,2}) = A_{1,1} + A_{1,2} + A_{2,1} + A_{2,2}$$

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Problem 3: Convolution terms

Summations

$$C_{k,l} = \sum_{i=k}^{k+M-1} \sum_{j=l}^{l+N-1} A_{i,j} \times B_{i-k+1, j-l+1}$$

$$C_{1,1} = \sum_{i=1}^{1+2-1} \sum_{j=1}^{1+2-1} A_{i,j} \times B_{i-1+1, j-1+1}$$

$M=N=2$
 $k=l=1$

$$= \sum_{i=1}^2 \sum_{j=1}^2 A_{i,j} \times B_{i,j}$$

$$= \sum_{i=1}^2 A_{i,1} \times B_{i,1} + A_{i,2} \times B_{i,2}$$

$$= A_{1,1} \times B_{1,1} + A_{1,2} \times B_{1,2} + A_{2,1} \times B_{2,1} + A_{2,2} \times B_{2,2}$$



Problem 3: Convolution terms

Convolution

Filter,
mask or
template

B

-1	2
-1	2

2×2

Input
image

A

4	4	7	9
4	3	8	9
3	5	9	9
3	6	10	9

4×4

Convolved
Image

C

6	23	21
9	26	19
16	27	17

3×3



Problem 3: Convolution terms

Convolution Size

Image size = $M_1 \times N_1$

←———— N_1 —————→

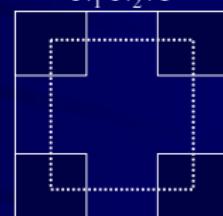
Mask size = $M_2 \times N_2$

← N_2 →

Convolution size =

$(M_1 - M_2 + 1) \times (N_1 - N_2 + 1)$

← $N_1 - N_2 + 1$ →



Typical Mask sizes

= $3 \times 3, 5 \times 5, 7 \times 7,$
 $9 \times 9, 11 \times 11$

What is the convolved image size for a 128×128 image and
 7×7 mask?



Problem 3: Convolution

The convolved image size for a 128×128 image and a 7×7 mask is 122×122 .



Problem 3: Convolution

Mask

-1	2	1
-1	2	1

Input image

4	4	7	1	9
4	3	8	2	9
3	5	9	8	9
3	6	9	8	9
3	6	10	9	9



Problem 3: Convolution

Output image

21	26	9
26	37	21
34	41	42
37	43	33





Outline

1 Problems

- Problem 1: Answer the questions
- Problem 2: Power law function
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- **Problem 4: Noise reduction**
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Problem 4: Noise reduction

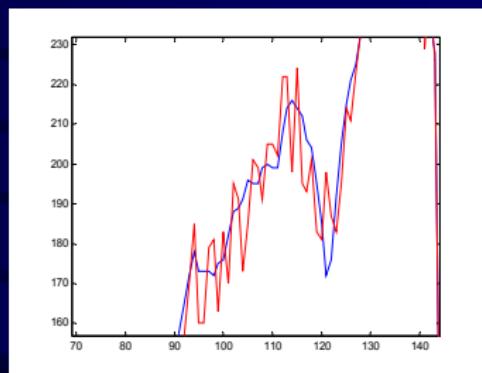
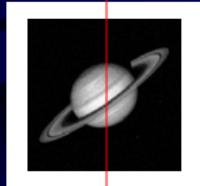
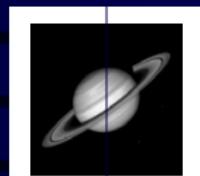
There are two methods of noise reduction: low pass filter and median filter (see P3: 37-46). Please give the noise reduction results of the input image in Q3 by using the two methods according to the conditions: a) the mask of low pass filter is

$$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}, \text{ and b) the mask of median filter is } \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$



Problem 4: Noise reduction terms

Noise Reduction



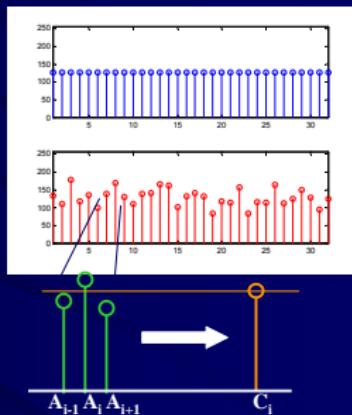
- Noise varies above and below uncorrupted image.



Problem 4: Noise reduction terms

Noise Reduction-1st principles

- How do we reduce noise?
- Consider a uniform 1-d image A and add noise.
- Focus on a pixel neighbourhood.
- Central pixel has been increased and neighbouring pixels have decreased.

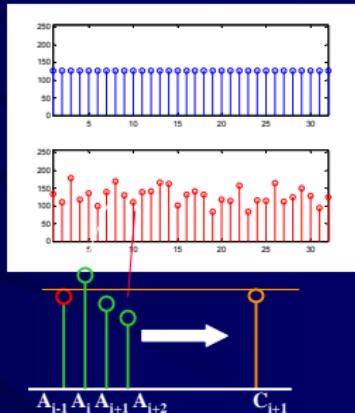


Problem 4: Noise reduction terms

Noise Reduction-1st principles

- Averaging ‘smoothes’ the noise fluctuations.
- Consider the next pixel A_{i+1}
- Repeat for remainder of pixels.

$$C_{i+1} = \frac{A_i + A_{i+1} + A_{i+2}}{3}$$





Problem 4: Noise reduction terms

Noise Reduction- Neighborhood Operations

- All pixels can be averaged by convolving I - d image A with mask \mathbf{B} to give enhanced image C .
- Weights of B must equal one when added together.

$$\mathbf{C} = \mathbf{A} * \mathbf{B}$$

$$\mathbf{B} = [B_1 \quad B_2 \quad B_3]$$

$$C_i = A_{i-1} \times B_1 + A_i \times B_2 + A_{i+1} \times B_3$$

$$B = \frac{1}{3} [1 \quad 1 \quad 1]$$

$$C_i = \frac{A_{i-1} + A_i + A_{i+1}}{3}$$

$$\mathbf{C} = \mathbf{A} * \mathbf{B}$$

$$\mathbf{B} = \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

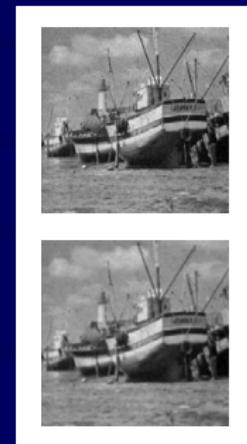
- Extend to two dimensions.



Problem 4: Noise reduction terms

Noise Reduction

- Technique relies on high frequency noise fluctuations being ‘blocked’ by filter. Hence, low-pass filter.
- Fine detail in image may also be smoothed.
- Balance between keeping image fine detail and reducing noise.
- Example:
 - Saturn image coarse detail
 - Boat image contains fine detail
 - Noise reduced but fine detail also smoothed

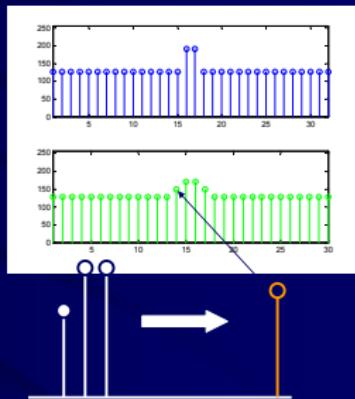




Problem 4: Noise reduction terms

Noise Reduction

- Consider a uniform 1-d image A with a step function.
- Step function corresponds to fine image detail such as an edge.
- Low-pass filter ‘blurs’ the edge.



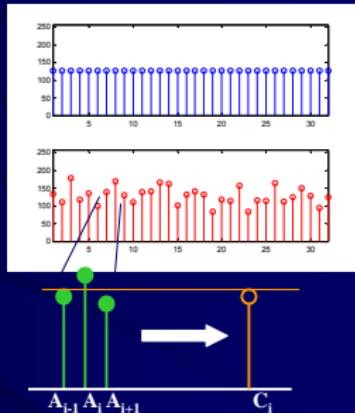


Problem 4: Noise reduction terms

Noise Reduction-1st principles

- How do we reduce noise without averaging?
- Consider a uniform 1-d image A and add noise.
- Focus on a pixel neighbourhood.
- Non-linear operator?

Median filter!





Problem 4: Noise reduction terms

Noise Reduction- Neighborhood Operations

- All pixels can be replaced by neighbourhood median by convolving 1-d image A with median filter B to give enhanced image C.
- Extend to two dimensions.

$$\begin{aligned}C &= \mathbf{A} * \mathbf{B} \\ \mathbf{B} &= [B_1 \quad B_2 \quad B_3] \\ C_i &= \text{median}\{A_{i-1} \times B_1, A_i \times B_2, A_{i+1} \times B_3\} \\ \mathbf{B} &= [1 \quad 1 \quad 1] \\ C_i &= \text{median}\{A_{i-1}, A_i, A_{i+1}\}\end{aligned}$$

$$\begin{aligned}C_{k,l} &= \underset{i=k:k+M-1, j=l:l+N-1}{\text{median}} \{A_{i,j} \times B_{i-k+1, j-l+1}\} \\ B_{i,j} &= 1 \text{ for all } i,j\end{aligned}$$

Problem 4: Noise reduction terms

Noise reduction

Original



Low-pass



Median

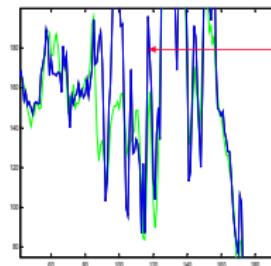




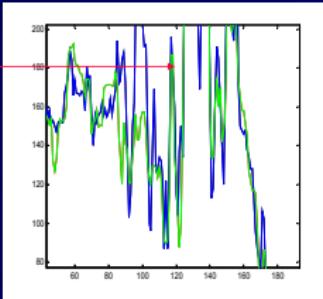
Problem 4: Noise reduction terms

Noise reduction

Low-pass



Median



- Low-pass: fine detail smoothed by averaging
- Median: fine detail passed by filter



Problem 4: Noise reduction

Input image

4	4	7	1	9
4	3	8	2	9
3	5	9	8	9
3	6	9	8	9
3	6	10	9	9

Low pass filter mask

$$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

Result a)

5	5	7
6	6	8
6	8	9



Problem 4: Noise reduction

Input image

4	4	7	1	9
4	3	8	2	9
3	5	9	8	9
3	6	9	8	9
3	6	10	9	9

Median filter mask

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

Result b)

4	5	8
5	8	9
6	8	9





Outline

1 Problems

- Problem 1: Answer the questions
- Problem 2: Power law function
- Problem 3: Convolution
- Problem 4: Noise reduction
- **Problem 5: Filter operations**



Problem 5: Filter operations

*In P3: 47-49 there are some examples of filter operations.
Please point out the difference of Low-pass filters, High-pass filters and Edge enhancement filters. Notice that the "0"-Sum mask for Edge enhancement filters and "1"-Sum for others.*



Problem 5: Filter operationsterms

Filter Operations

1 Low-Pass Filter

- Class: Image Enhancement/Restoration
- Implementation: Pixel group process
and smooth an image

1 1 1

1 1 1

1 1 1

/9

1 1 1

1 2 1

1 1 1

/10

1 2 1

2 4 2

1 2 1

/16



Problem 5: Filter operationsterms

Filter Operations

2 High-Pass Filter

- Implementation: Pixel group process and sharpen an image

-1 -1 -1

0 -1 0

1 -2 1

-1 9 -1

-1 5 -1

-2 5 -2

-1 -1 -1

0 -1 0

1 -2 1

3 Sobel Edge Enhancement

- Implementation: Edge extraction

-1 0 1

-1 -2 -1

-2 0 2

0 0 0

-1 0 1

1 2 1

Vertical mask

Horizontal mask



Problem 5: Filter operationsterms

Filter Operations

4 Shift and Difference Edge Enhancement

- Implementation: Vertical, Horizontal and Diagonal Edge extraction

0 0 0

-1 1 0

0 0 0

Vertical

0 -1 0

0 1 0

0 0 0

Horizontal

-1 0 0

0 1 0

0 0 0

Diagonal

5 Laplacian Edge Enhancement

- Implementation: All Edge extraction

-1 -1 -1

-1 8 -1

-1 -1 -1

0 -1 0

-1 4 -1

0 -1 0

1 -2 1

-2 4 -2

1 -2 1



Problems

Any questions?