

Digital Image Processing

COSC 6380/4393

Midterm Review

Oct 26th, 2023

Mid Term Exam

- Syllabus:
 - Introduction
 - Binary Image Processing
 - Point Operations
 - Discrete Fourier Transform
 - Spatial Filtering

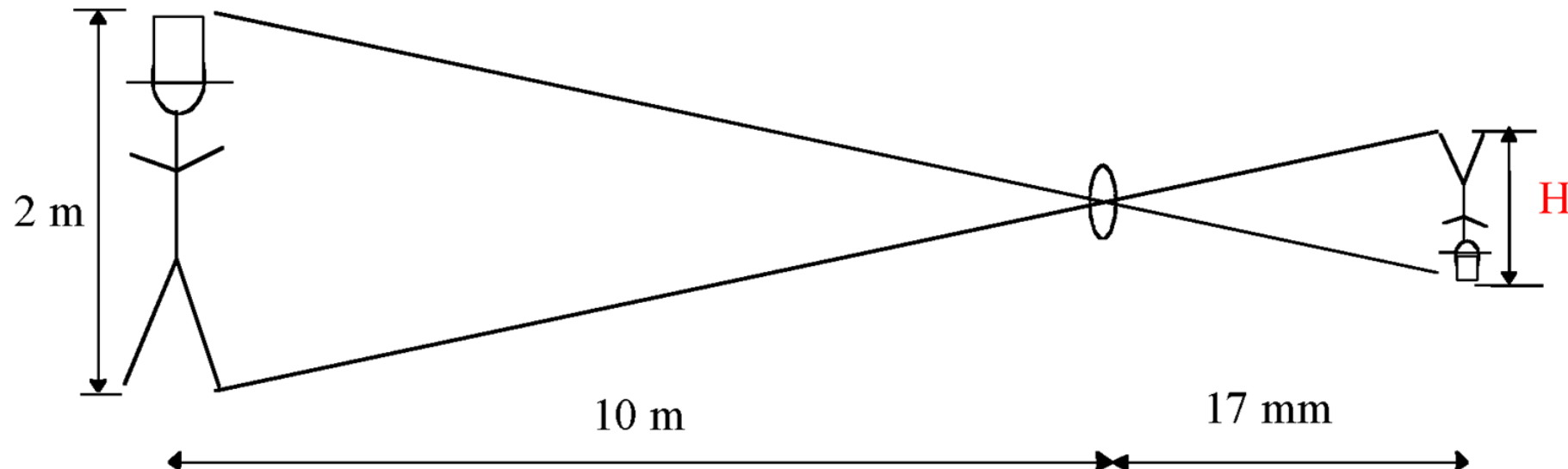
Image Formation

Image Formation: (Projection)

1. A person is standing M meters in front of a camera
 - h is the height of the person
 - X is the focal length of the camera
 - H is the height of the projection of the person on the imaging plane. What is H ? (Show steps)
2. A sphere/square is placed at a distance of M meters (m) in front of a camera
 - a/v is the area/volume of the object
 - X is the focal length of the camera
 - A/V is the area of the projection of the object on the imaging plane. What is A/V ? (Show steps)

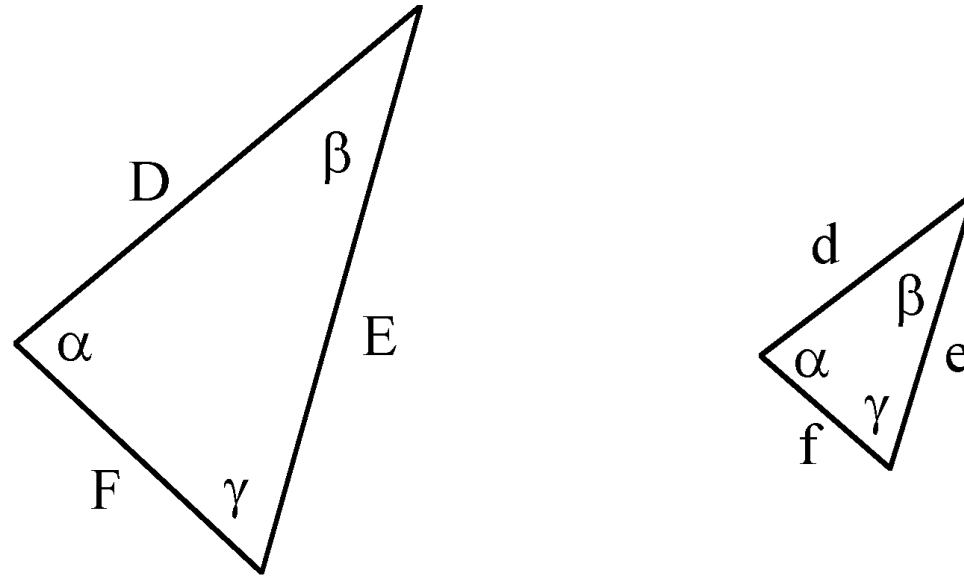
Example: Solution

- There is a man standing 10 meters (m) in front of you
- He is 2 m tall
- The focal length of your eye is about 17 mm
- Question: What is the height H of his image on your retina?



SIMILAR TRIANGLES

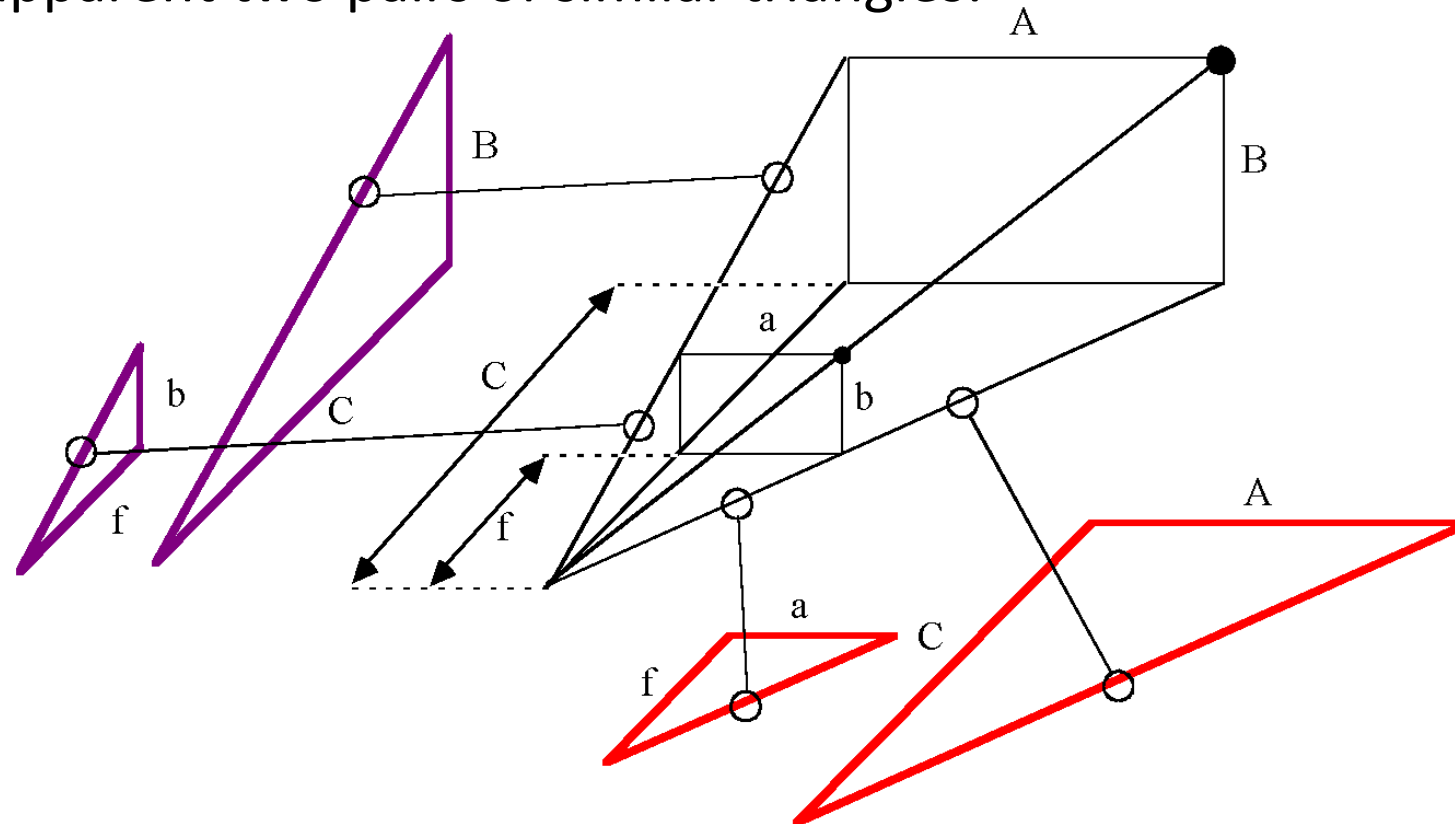
- Similar Triangles Theorem - Similar triangles have their side lengths in the same proportions.



$$\frac{D}{E} = \frac{d}{e} \qquad \frac{E}{F} = \frac{e}{f} \qquad \frac{F}{D} = \frac{f}{d}$$

SOLVING PERSPECTIVE PROJECTION

- Using similar triangles we can solve for the relationship between 3-D coordinates in space and 2-D image coordinates
- Redraw the imaging geometry once more, this time making apparent two pairs of similar triangles:



SOLVING PERSPECTIVE PROJECTION

- By the Similar Triangles Theorem, we conclude that

$$\frac{a}{f} = \frac{A}{C} \quad \text{and} \quad \frac{b}{f} = \frac{B}{C}$$

OR

$$(a, b) \stackrel{f}{\underset{C}{=}} \cdot (A, B) = (fA/C, fB/C)$$

PERSPECTIVE PROJECTION EQUATION

- Thus the following relationship holds between 3-D space coordinates (X, Y, Z) and 2-D image coordinates (x, y) :

$$(x, y) = \frac{f}{Z} \cdot (X, Y)$$

where f = focal length.

- The ratio f/Z is the magnification factor, which varies with the range Z from the lens center to the object plane.

ANSWER

- By similar triangles,

$$\frac{2 \text{ m}}{10 \text{ m}} = \frac{H}{17 \text{ mm}}$$

$$\underline{H = 3.4 \text{ mm}}$$

Alternatives: Given a circle or square of known area, parallel to the imaging plane.
How do you compute area of the projected image?

Bilinear Interpolation

Given the location of four pixels Q_{11} , Q_{12} , Q_{21} , Q_{22} and their intensity values I_{11} , I_{12} , I_{21} , I_{22} .
Assuming that Q_{11} , Q_{12} , Q_{21} , Q_{22} are the nearest pixels to P

Estimate the intensity value of pixel located at P using bilinear interpolation?

Bi-Linear Interpolation(2D)

$$Q_{11} = (x_1, y_1),$$

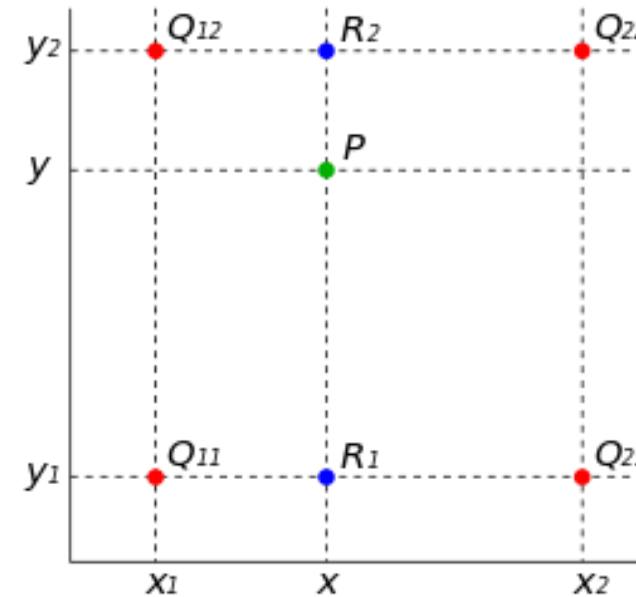
$$Q_{12} = (x_1, y_2),$$

$$Q_{21} = (x_2, y_1),$$

$$\text{and } Q_{22} = (x_2, y_2)$$

$f(Q_i) \rightarrow \text{intensity at } Q_i$

Find the value at P



Example: Linear Interpolation



Solve for I

$$I = \frac{I_1(x_2 - x)}{(x_2 - x_1)} + \frac{I_2(x - x_1)}{(x_2 - x_1)}$$
$$I = \frac{10(1 - 0.3)}{(1 - 0)} + \frac{15(0.3 - 0)}{(1 - 0)}$$
$$I = 7 + 4.5 = 11.5$$

Example

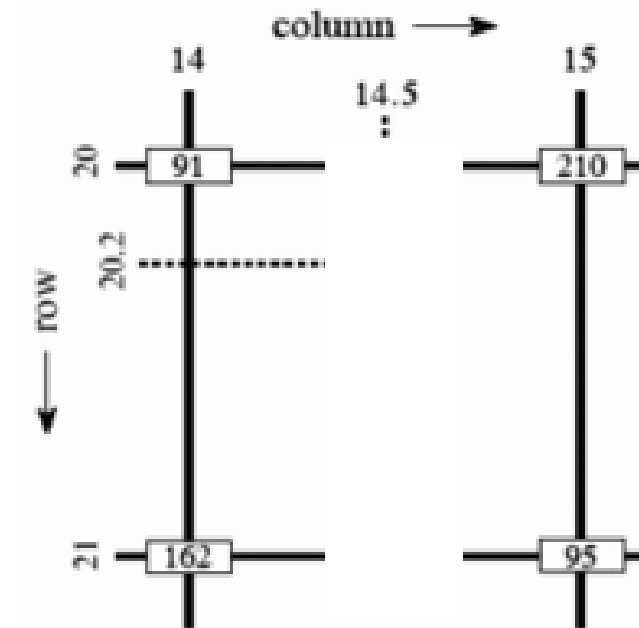
$$I(21,14) = 162,$$

$$I(21,15) = 95,$$

$$I(20,14) = 91,$$

$$I(20,15) = 210$$

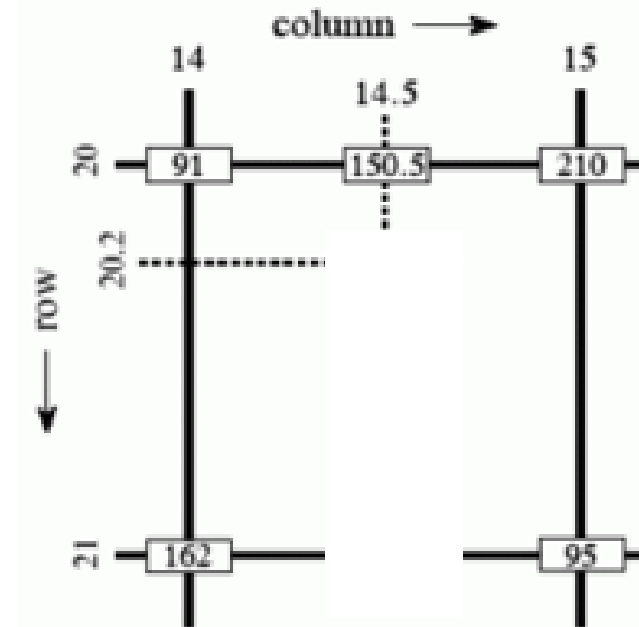
$$I(20.2, 14.5) = ?$$



Example

$$\begin{aligned} I(21,14) &= 162, \\ I(21,15) &= 95, \\ I(20,14) &= 91, \\ I(20,15) &= 210 \\ I(20.2, 14.5) &= ? \end{aligned}$$

$$I_{20,14.5} = \frac{15-14.5}{15-14} \cdot 91 + \frac{14.5-14}{15-14} \cdot 210 = 150.5,$$



Example

$$I(21,14) = 162,$$

$$I(21,15) = 95,$$

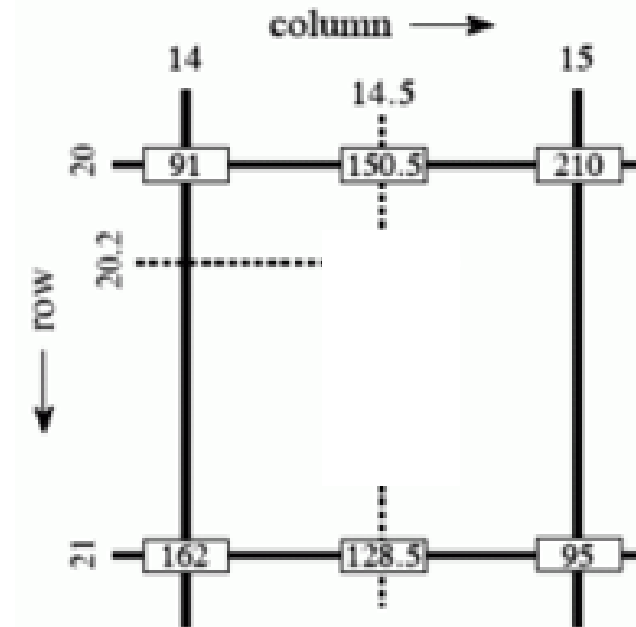
$$I(20,14) = 91,$$

$$I(20,15) = 210$$

$$I(20.2, 14.5) = ?$$

$$I_{20,14.5} = \frac{15-14.5}{15-14} \cdot 91 + \frac{14.5-14}{15-14} \cdot 210 = 150.5,$$

$$I_{21,14.5} = \frac{15-14.5}{15-14} \cdot 162 + \frac{14.5-14}{15-14} \cdot 95 = 128.5,$$



Example

$$I(21,14) = 162,$$

$$I(21,15) = 95,$$

$$I(20,14) = 91,$$

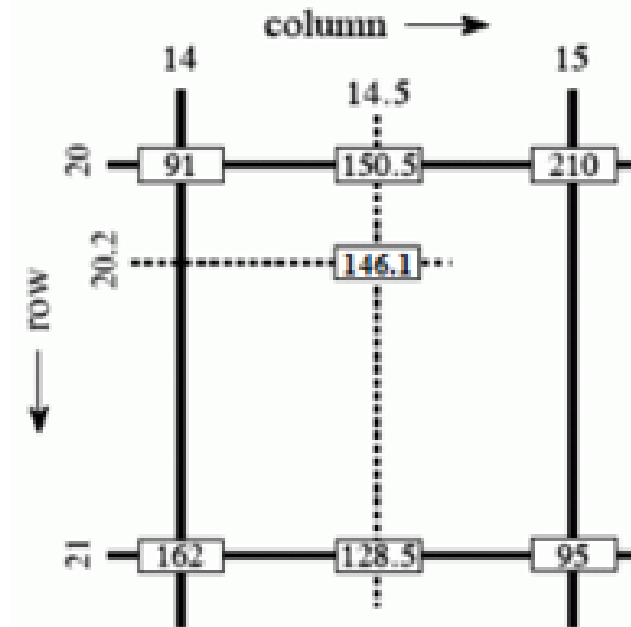
$$I(20,15) = 210$$

$$I(20.2, 14.5) = ?$$

$$I_{20,14.5} = \frac{15-14.5}{15-14} \cdot 91 + \frac{14.5-14}{15-14} \cdot 210 = 150.5,$$

$$I_{21,14.5} = \frac{15-14.5}{15-14} \cdot 162 + \frac{14.5-14}{15-14} \cdot 95 = 128.5,$$

$$I_{20.2,14.5} = \frac{21-20.2}{21-20} \cdot 150.5 + \frac{20.2-20}{21-20} \cdot 128.5 = 146.1.$$



Binary Image Processing

Binary Image Logical Operations

- Given an acquired binary images I and a model binary Image M below. Generate a third binary image D representing the unmatched pixels in the acquired image compared to the model image.
- Since binary operations are quicker, you are allowed to use only binary operators (And, OR and NOT) or a combination of these on binary image M and I to accomplish this task.

0	1
1	0

 M

0	1
0	1

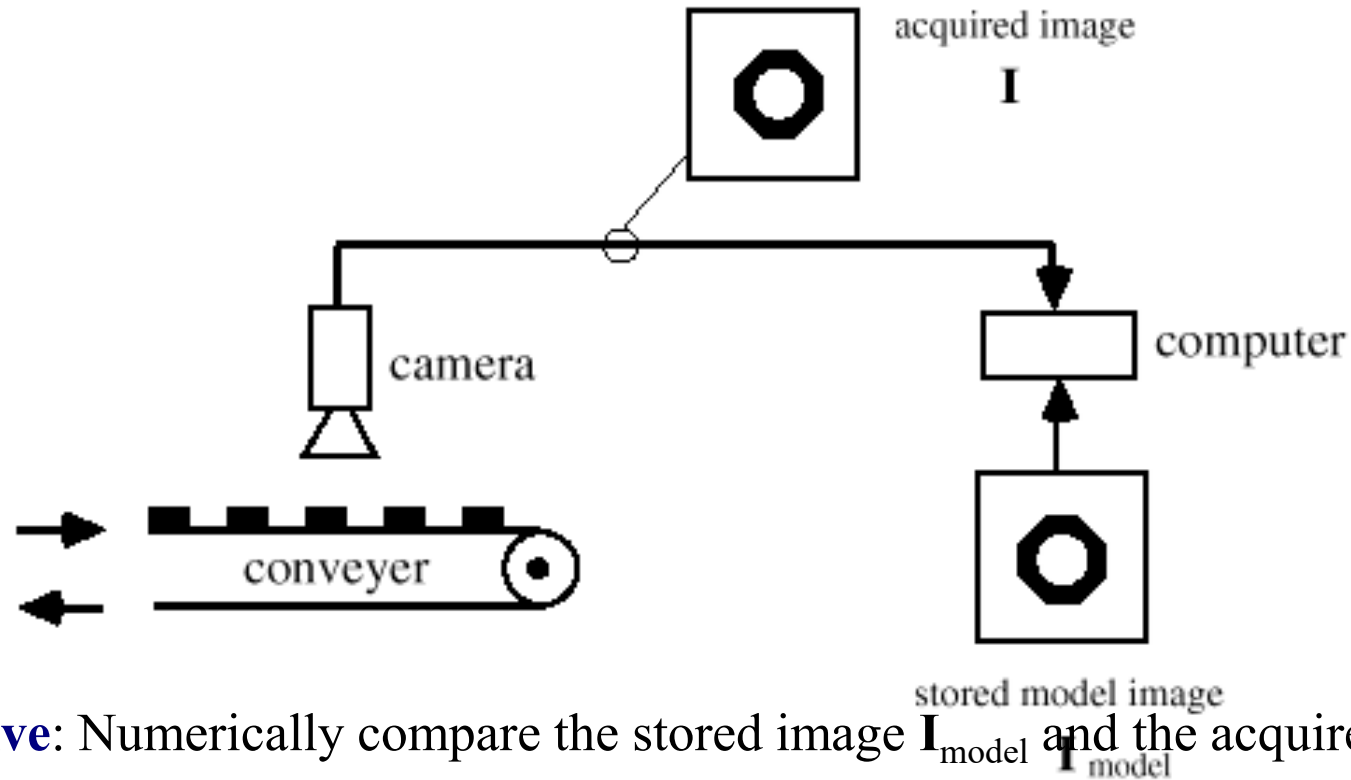
 I

0	0
1	1

 D

3. Image difference

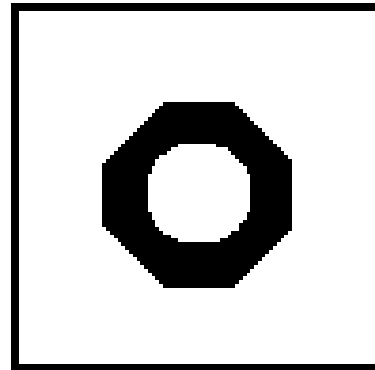
- An assembly-line image inspection system. Similar to many marketed by industry:



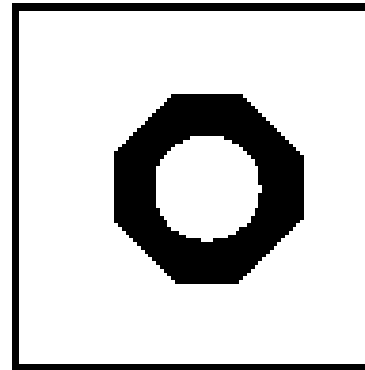
- Objective:** Numerically compare the stored image I_{model} and the acquired image I

EXAMPLE

- Observe that the object in **I** has been shifted very slightly



I model

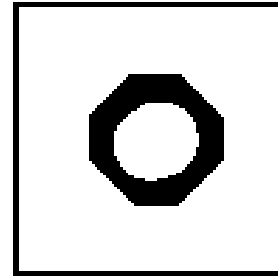


I



Logical AND

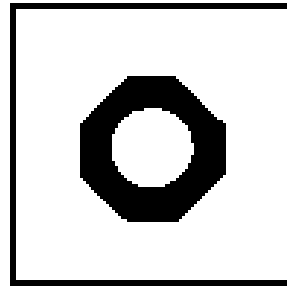
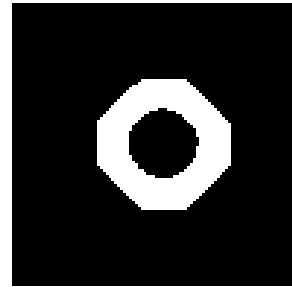
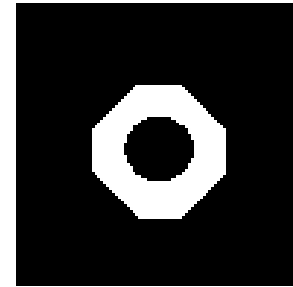
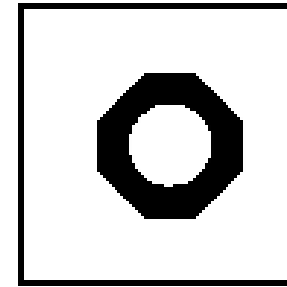
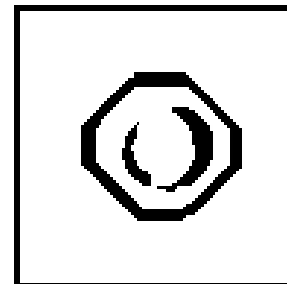
- The logical AND conveys the **overlap**



$$\mathbf{I}_{\text{model}} \wedge \mathbf{I}$$

- A measurement of the **displacement** is given by:
- $\text{XOR}(\mathbf{I}, \mathbf{I}_{\text{model}}) = \text{OR} \{ \text{AND}[\mathbf{I}_{\text{model}}, \text{NOT}(\mathbf{I})], \text{AND}[\text{NOT}(\mathbf{I}_{\text{model}}), \mathbf{I}] \}$

DISPLACEMENT

 I_{model}  $\text{NOT}(I)$  $\text{NOT}(I_{\text{model}})$  I  $\text{XOR}(I, I_{\text{model}})$

4. Morphological Operations

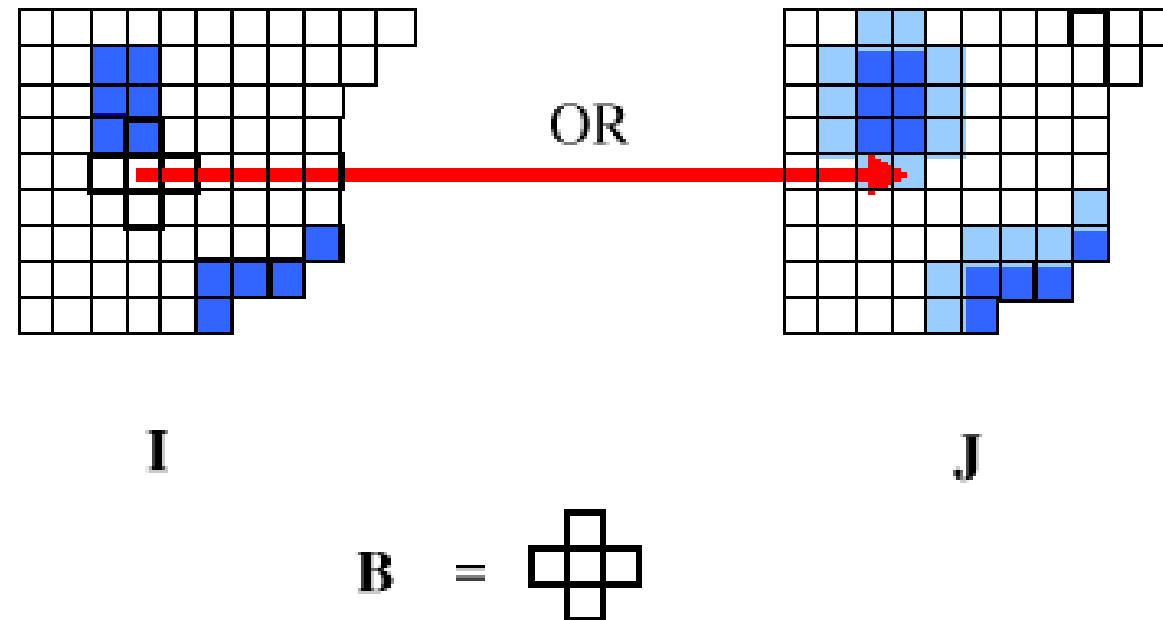
1. Perform a morphological operation using a window of your choice to remove the small object represented by 1 at pixel (2,4) in the image below.

0	0	0	0	0
1	1	0	1	0
1	1	0	0	0
0	1	0	0	0
0	0	0	0	0

Operations to cover holes, remove peninsula, etc.?

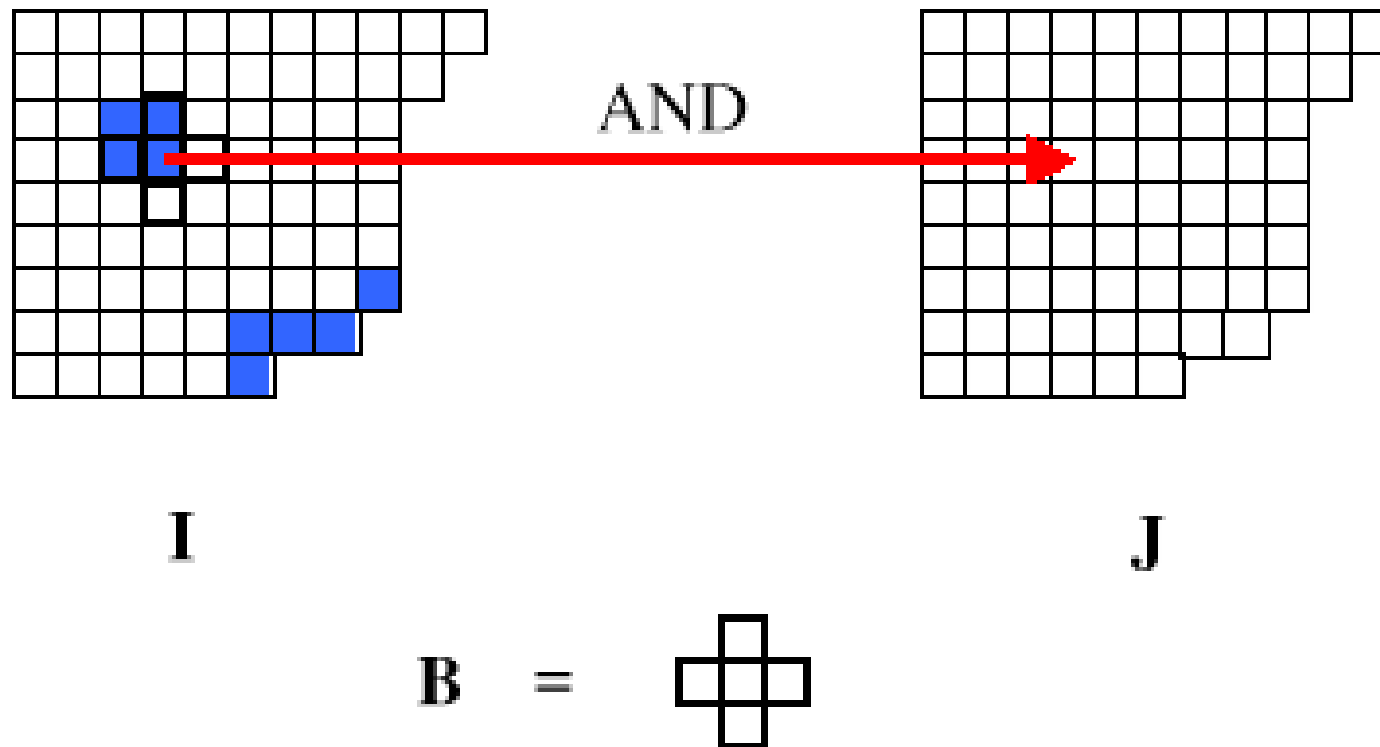
DILATION

- So-called because this operation **increases** the size of BLACK objects in a binary image
- Local Computation: $J = \text{DILATE}(I, B)$



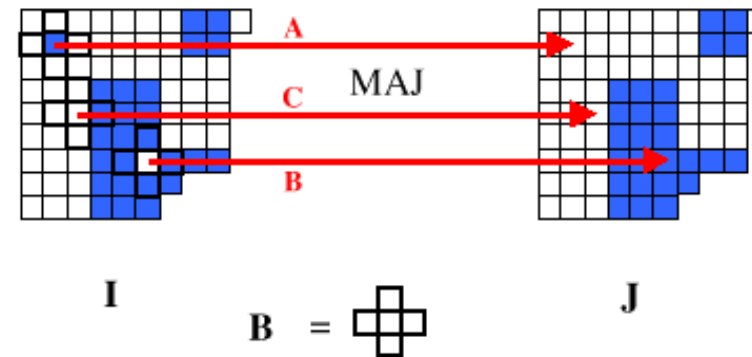
EROSION

- So-called because this operation **decreases** the size of BLACK objects in a binary image
- Local Computation: $J = \text{ERODE}(I, B)$



MEDIAN

- Actually **majority**. A special case of the gray-level **median filter**
- Possesses qualitative attributes of both dilation and erosion, but does not generally change the **size** of objects or background
- Local Computation: $J = \text{MEDIAN}(I, B)$

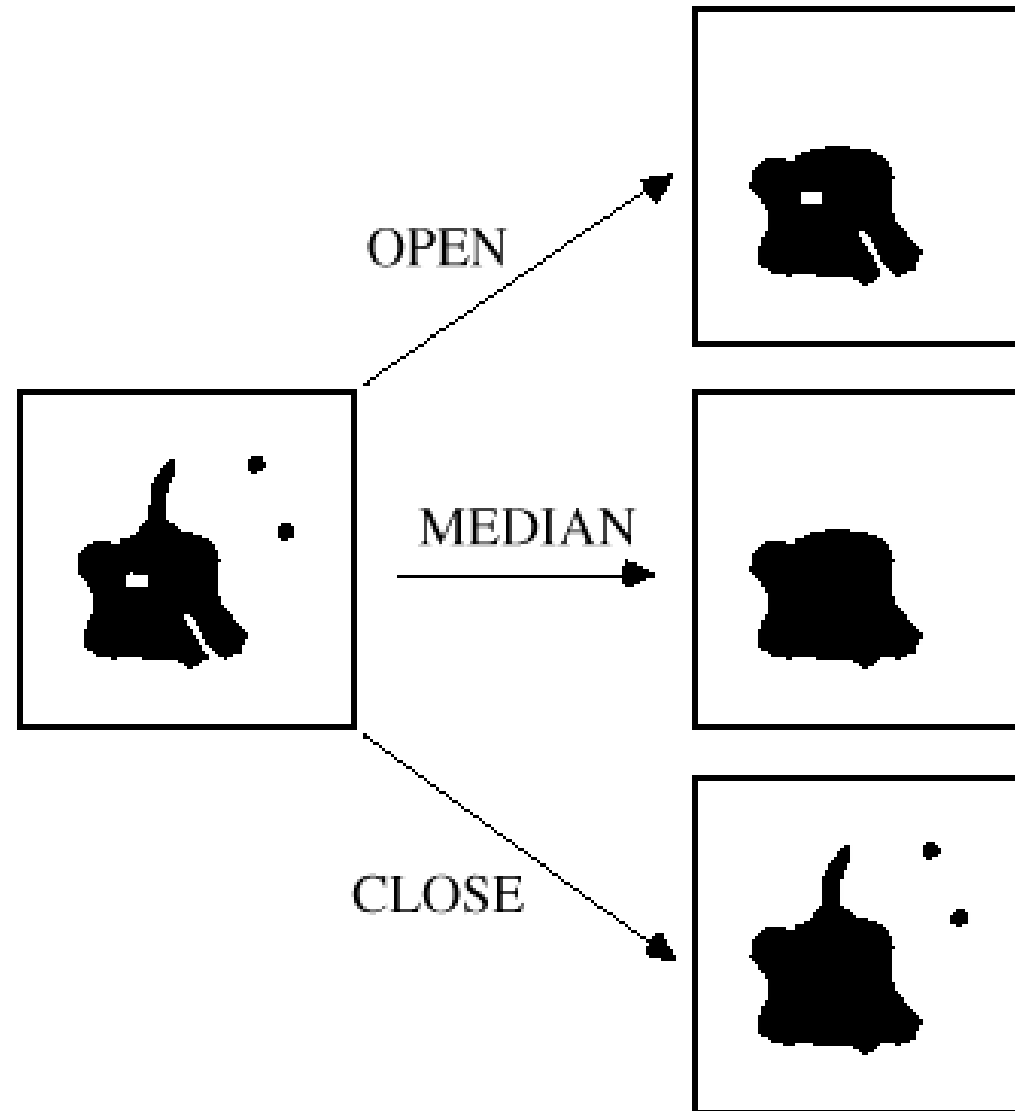


- The median removed the small **object** A and the small **hole** B , but did not change the boundary (**size**) of the larger region C

OPENing and CLOSing

- We can define **new** morphological operations by performing the basic ones in sequence
- Given an image **I** and window **B**, define
$$\text{OPEN}(\mathbf{I}, \mathbf{B}) = \text{DILATE} [\text{ERODE}(\mathbf{I}, \mathbf{B}), \mathbf{B}]$$
$$\text{CLOSE}(\mathbf{I}, \mathbf{B}) = \text{ERODE} [\text{DILATE}(\mathbf{I}, \mathbf{B}), \mathbf{B}]$$
- In other words,
- OPEN = erosion (by **B**) followed by dilation (by **B**)
- CLOSE = dilation (by **B**) followed by erosion (by **B**)

EXAMPLES



4. Morphological Operations

1. Perform a morphological operation using a window of your choice to remove the small object represented by 1 at pixel (1,3) in the image below.

0	0	0	0	0
1	1	0	1	0
1	1	0	0	0
0	1	0	0	0
0	0	0	0	0

1. Operations to cover holes, remove peninsula?

Solution

- A possible solution can be using the median operation with a filter (3,1)

0	0	0	0	0
1	1	0	1	0
1	1	0	0	0
0	1	0	0	0
0	0	0	0	0

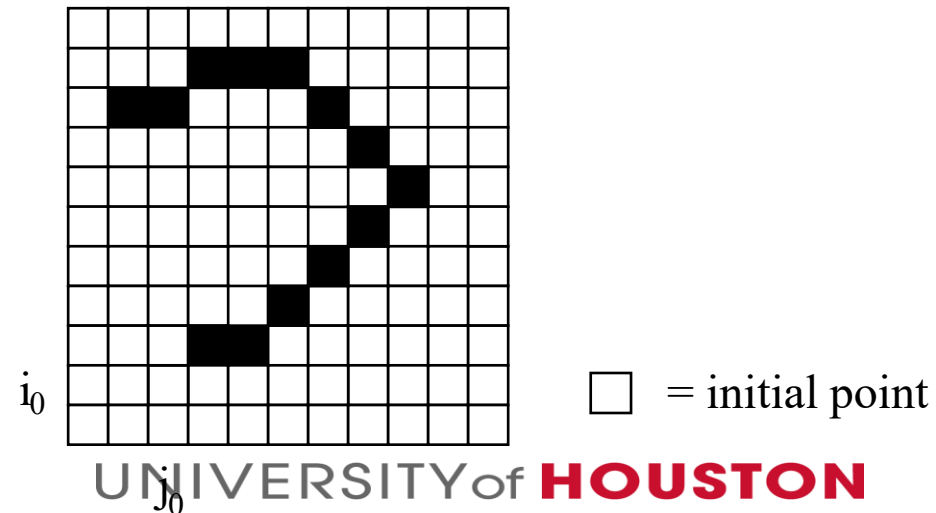
0	0	0	0	0
1	1	0	0	0
1	1	0	0	0
0	1	0	0	0
0	0	0	0	0

5. Compression

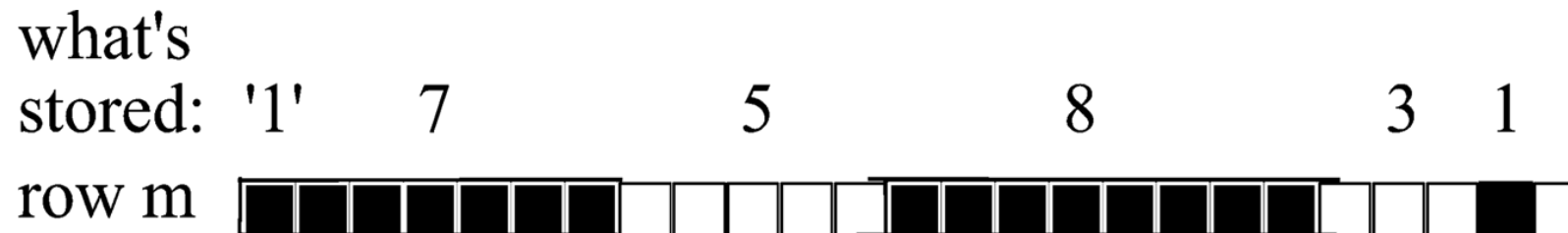
1. Perform compression to represent sequence of pixels in the binary image below

what's
stored: '1' 7 5 8 3 1
row m 

1. Perform compression to represent the contour in the binary image using chain codes



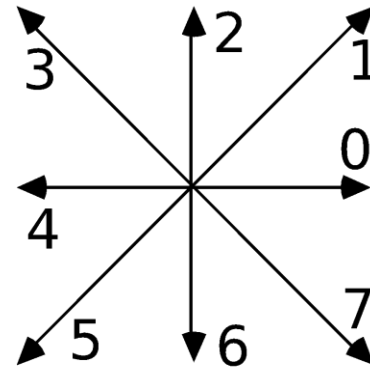
Run Length Encoding



Code: 175831

CHAIN CODE

- We use the following 8-neighbor direction codes:

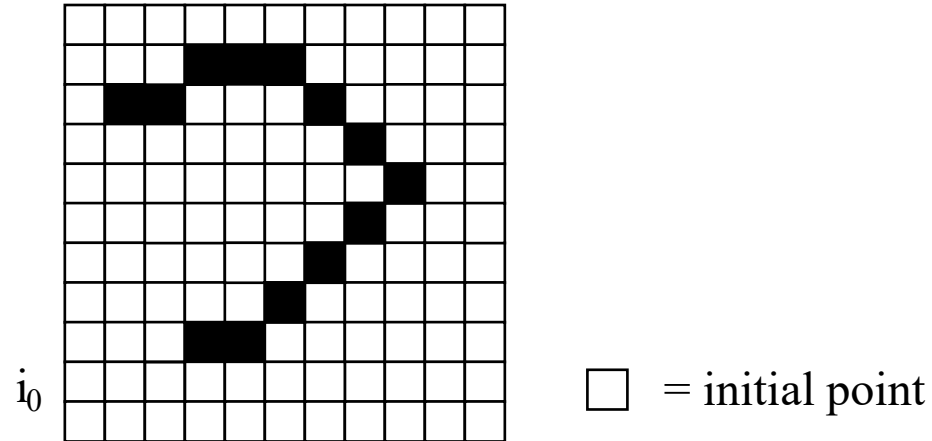


- Since the numbers 0, 1, 2, 3, 4, 5, 6, 7 can be coded by their 3-bit binary equivalents:

000, 001, 010, 011, 100, 101, 110, 111

the location of each point on the contour **after** the initial point can be coded by 3 bits.

EXAMPLE



- Its chain code: (after recording the initial coordinate (i_0, j_0))

1, 0, 1, 1, 1, 1, 3, 3, 3, 4, 4, 5, 4

=

001, 000, 001, 001, 001, 001, 011, 011, 011, 100, 100,
101, 100

Point Operations

Linear Point Operations

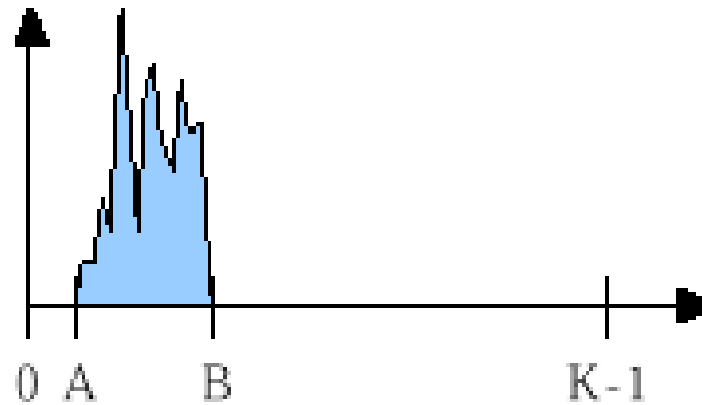
1. Perform a full contrast stretch on the image below assuming that each dynamic range of the image is 0-15
2. Perform histogram flattening, shaping?

$$\mathbf{I} =$$

1	1	3	4
2	5	3	2
8	1	8	2
4	5	3	11

6. Full-Scale Contrast Stretch

- The **most common** linear point operation. Suppose **I** has a compressed histogram:



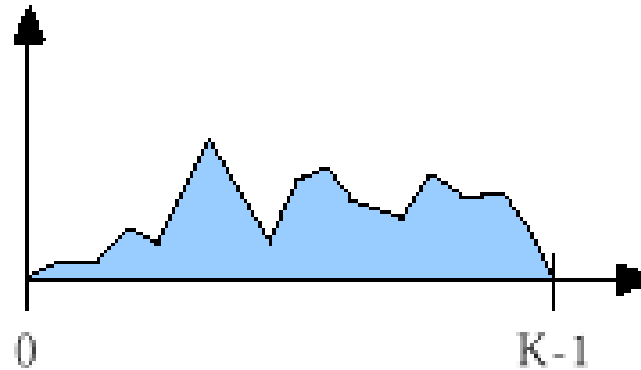
- Let A and B be the min and max gray levels in **I**
- Define

$$J(i, j) = P \cdot I(i, j) + L$$

- such that $P \cdot A + L = 0$ and $P \cdot B + L = (K-1)$

Full-Scale Contrast Stretch

- The result of solving these **2 equations in 2 unknowns** (P, L) is an image **J** with a full-range histogram:



- The solution to the above equations is

$$P = \left\lfloor \frac{K-1}{B-A} \right\rfloor \quad \text{and} \quad L = -A \left\lfloor \frac{K-1}{B-A} \right\rfloor$$

or

$$J(i, j) = \left\lfloor \frac{K-1}{B-A} \right\rfloor [I(i, j) - A]$$

- $B = 11$
- $A = 1$
- $B - A = 10$
- $(K - 1) = 15$

- If I_c is the image with full dynamic range.

- $I_c[0,0] = INT\left(\frac{15}{10} (I[0,0] - 1) + 0.5\right)$

$$= INT\left(\frac{15}{10} (1 - 1) + 0.5\right) = 0$$

- $I_c[0,2] = INT\left(\frac{15}{10} (I[0,2] - 1) + 0.5\right)$

$$= INT\left(\frac{15}{10} (3 - 1) + 0.5\right) = 3$$

...

$\mathbf{I} =$

1	1	3	4
2	5	3	2
8	1	8	2
4	5	3	11

- $B = 11$
- $A = 1$
- $B - A = 10$
- $(K - 1) = 15$

- If I_c is the image with full dynamic range.

- $I_c[0,0] = INT\left(\frac{15}{10} (I[0,0] - 1) + 0.5\right)$

$$= INT\left(\frac{15}{10} (1 - 1) + 0.5\right) = 0$$

- $I_c[0,2] = INT\left(\frac{15}{10} (I[0,2] - 1) + 0.5\right)$

$$= INT\left(\frac{15}{10} (3 - 1) + 0.5\right) = 3$$

...

$$\mathbf{I} =$$

1	1	3	4
2	5	3	2
8	1	8	2
4	5	3	11

Result:

0	0	3	5
2	6	3	2
11	0	11	2
5	6	3	15

Histogram Flattening

- Given a 4 x 4 image I with gray-level range $\{0, \dots, 15\}$ ($K-1 = 15$):

$$I = \begin{array}{|c|c|c|c|} \hline 1 & 1 & 3 & 4 \\ \hline 2 & 5 & 3 & 2 \\ \hline 8 & 1 & 8 & 2 \\ \hline 4 & 5 & 3 & 11 \\ \hline \end{array}$$

- It's histogram is

k	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
H(k)	0	3	3	3	2	2	0	0	2	0	0	1	0	0	0	0



Histogram Flattening

- The normalized histogram is

k	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
p(k)	0	$\frac{3}{16}$	$\frac{3}{16}$	$\frac{3}{16}$	$\frac{2}{16}$	$\frac{2}{16}$	0	0	$\frac{2}{16}$	0	0	$\frac{1}{16}$	0	0	0	0

- From which we can compute the intermediate image **J₁** and finally the "flattened" image **J**:

J₁ =

$3/16$	$3/16$	$9/16$	$11/16$
$6/16$	$13/16$	$9/16$	$6/16$
$15/16$	$3/16$	$15/16$	$6/16$
$11/16$	$13/16$	$9/16$	$16/16$

J =

3	3	8	10
6	12	8	6
14	3	14	6
10	12	8	15

Histogram Shaping

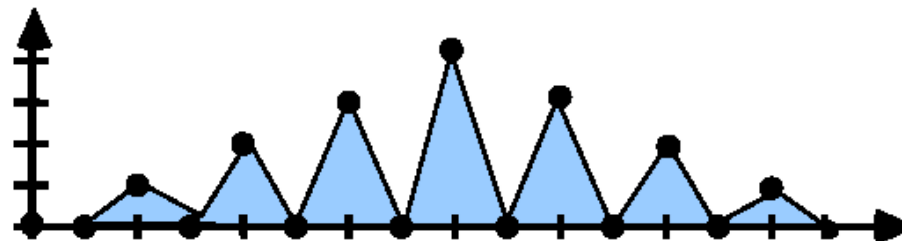
- Consider the same image as in the last example. We had

$$\mathbf{I} = \begin{array}{|c|c|c|c|} \hline 1 & 1 & 3 & 4 \\ \hline 2 & 5 & 3 & 2 \\ \hline 8 & 1 & 8 & 2 \\ \hline 4 & 5 & 3 & 11 \\ \hline \end{array}$$

$$\mathbf{J}_1 = \begin{array}{|c|c|c|c|} \hline 3/16 & 3/16 & 9/16 & 11/16 \\ \hline 6/16 & 13/16 & 9/16 & 6/16 \\ \hline 15/16 & 3/16 & 15/16 & 6/16 \\ \hline 11/16 & 13/16 & 9/16 & 16/16 \\ \hline \end{array}$$

- Fit this t

k	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
$H_J(k)$	0	0	1	0	2	0	3	0	4	0	3	0	2	0	1	0
$p_J(k)$	0	0	$\frac{1}{16}$	0	$\frac{2}{16}$	0	$\frac{3}{16}$	0	$\frac{4}{16}$	0	$\frac{3}{16}$	0	$\frac{2}{16}$	0	$\frac{1}{16}$	0



Histogram Shaping

- Here's the cumulative (summed) probabilities associated with it:

n	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
$P_J(n)$	0	0	$\frac{1}{16}$	$\frac{1}{16}$	$\frac{3}{16}$	$\frac{3}{16}$	$\frac{6}{16}$	$\frac{6}{16}$	$\frac{10}{16}$	$\frac{10}{16}$	$\frac{13}{16}$	$\frac{13}{16}$	$\frac{15}{16}$	$\frac{15}{16}$	$\frac{16}{16}$	$\frac{16}{16}$

- Careful** visual inspection of J_1 let's us form the new image:

$\mathbf{J} =$

4	4	8	10
6	10	8	6
12	4	12	6
10	10	8	14

HISTOGRAM MATCHING

- Just a special case of histogram shaping.
- Difference: the histogram of the original image I is matched to that of another image I' .
- Otherwise the procedure is identical, once the cumulative probabilities are computed for the model image I' .
- Useful application: **Comparing** similar images of the same scene obtained under different conditions (e.g., lighting, time of day). Extends the concept of equalizing AOD described earlier.