

## **DMET 901 – Computer Vision**

# ***Image Transformation and Filtering***

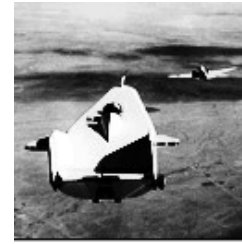
Seif Eldawlatly

# Introduction

- Image preprocessing has two goals
  - Suppress information that is not relevant (Noise,...)
  - Enhance information that is relevant (Edges, ....)

- Three topics

- Transformations



- Noise Filtering



- Edge Detection



# Transformations

- Gray-scale Transformations
  - A transform  $T$  of the original brightness  $p$  from the scale  $[p_0, p_k]$  into the brightness  $q$  from a new scale  $[q_0, q_k]$ 
$$q = T(p)$$

- Example: Increase the contrast



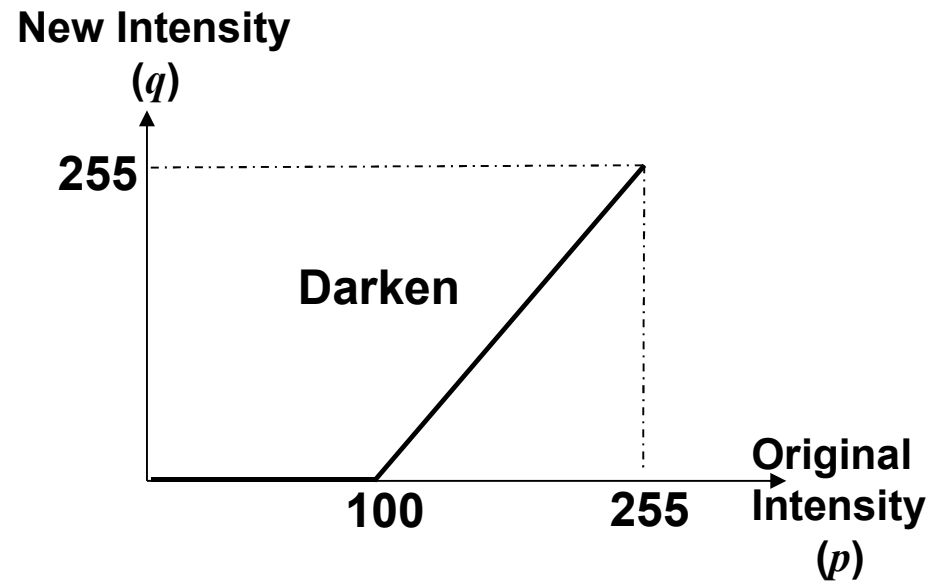
**Original Image**



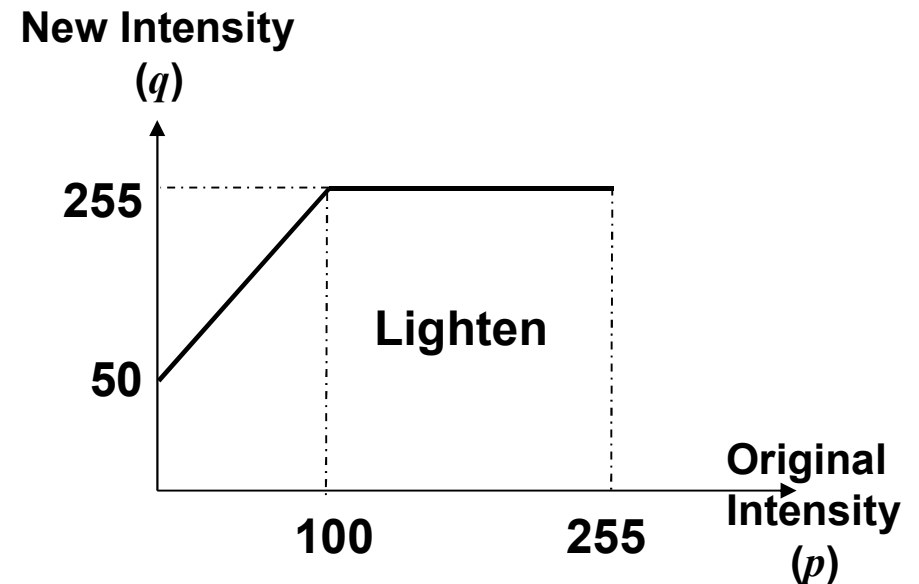
**Increase Contrast**

# Gray-scale Transformation

- Examples of simple operations on gray-scale images:  
*Change Brightness*



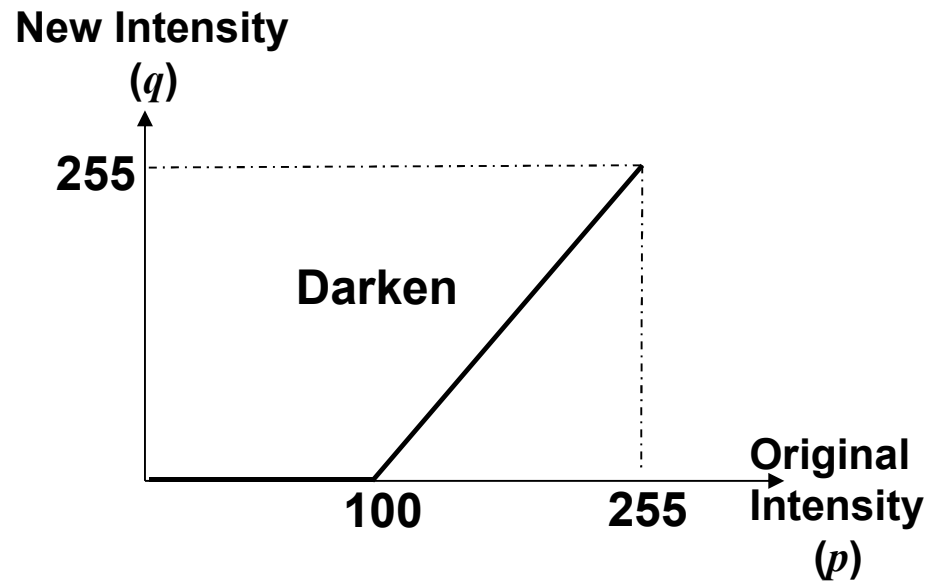
$$q = T(p) = \begin{cases} 0, & \text{if } p \leq 100 \\ \frac{255}{155}(p - 100), & \text{if } p > 100 \end{cases}$$



$$q = T(p) = \begin{cases} \frac{1}{100}(205p - 5000), & \text{if } p \leq 100 \\ 255, & \text{if } p > 100 \end{cases}$$

# Gray-scale Transformation

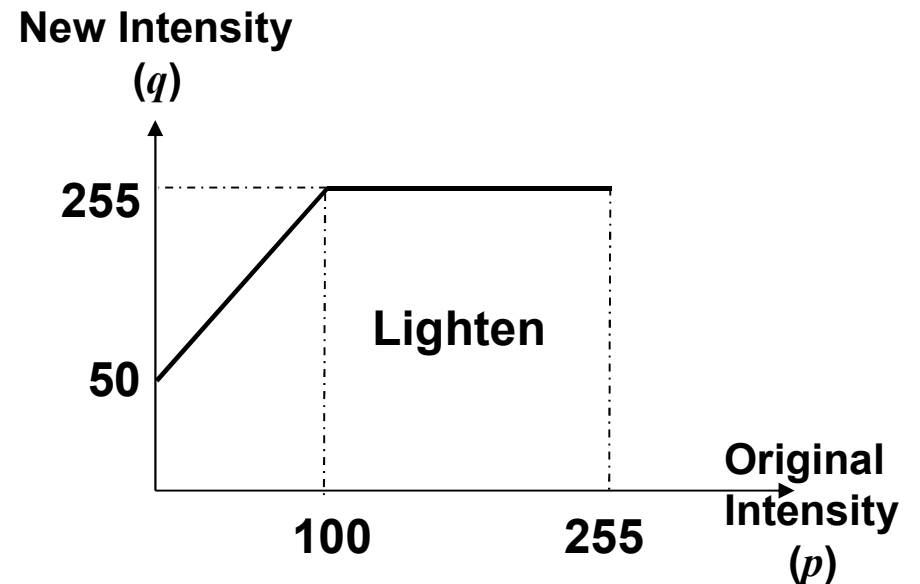
- Examples of simple operations on gray-scale images:  
*Change Brightness*



Original



Darkened

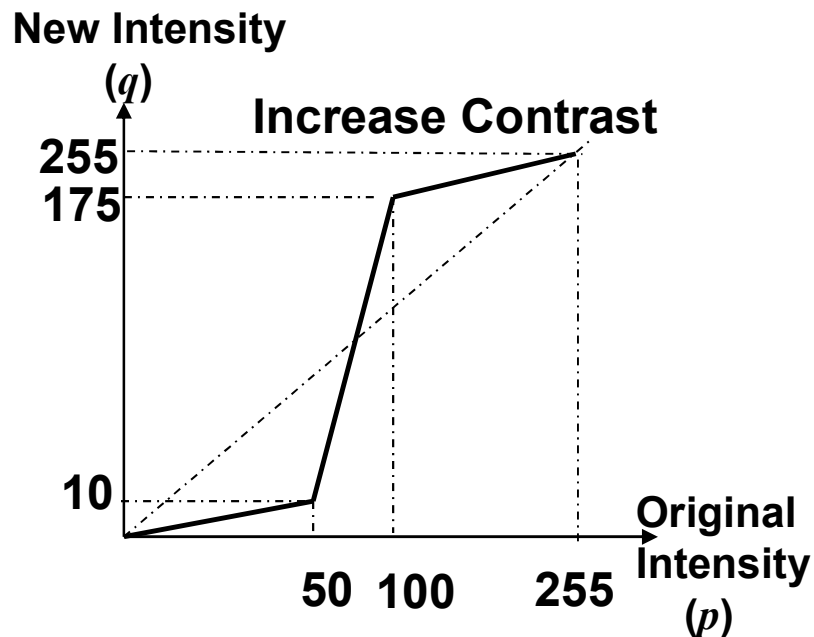


Lightened

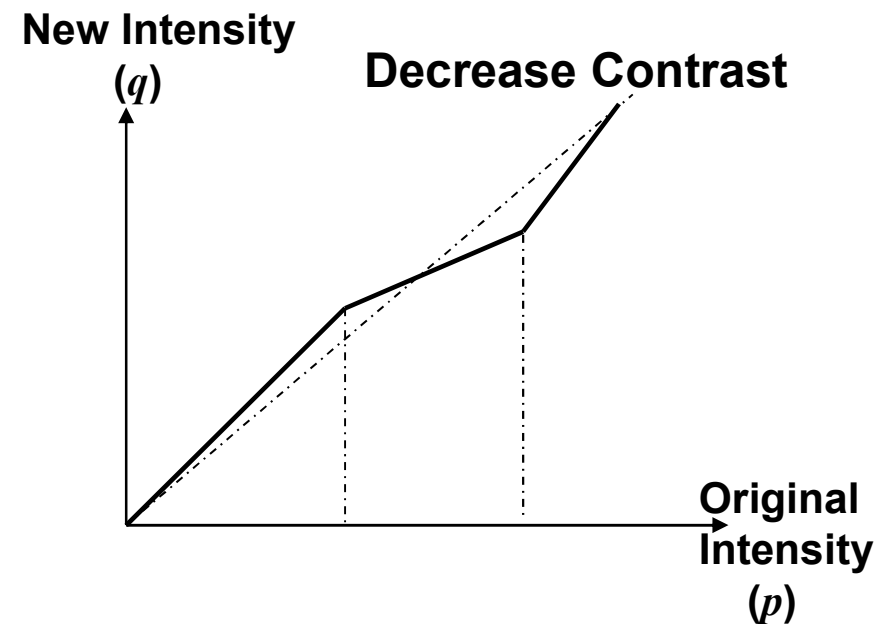


# Gray-scale Transformation

- Examples of simple operations on gray-scale images:  
*Change Contrast*



**Bright becomes brighter,  
Dark becomes darker**



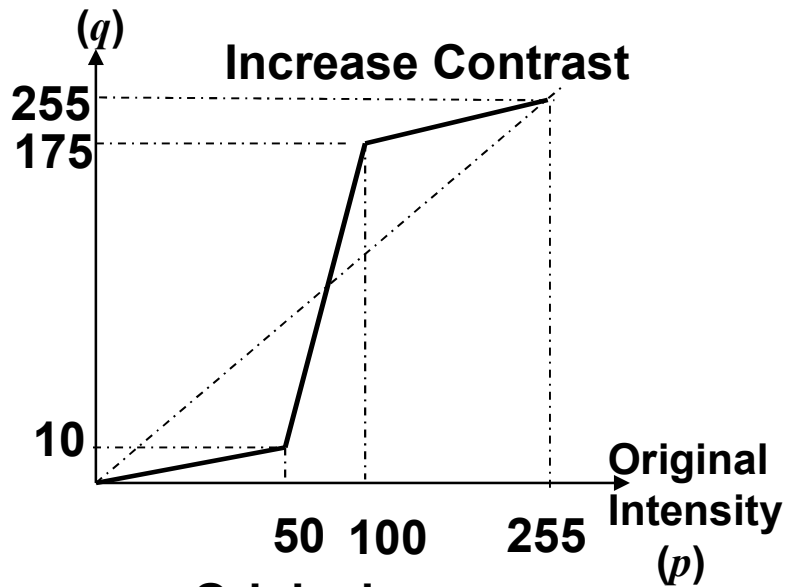
**Bright becomes darker,  
Dark becomes brighter**

# Gray-scale Transformation

- Examples of simple operations on gray-scale images:

## *Change Contrast*

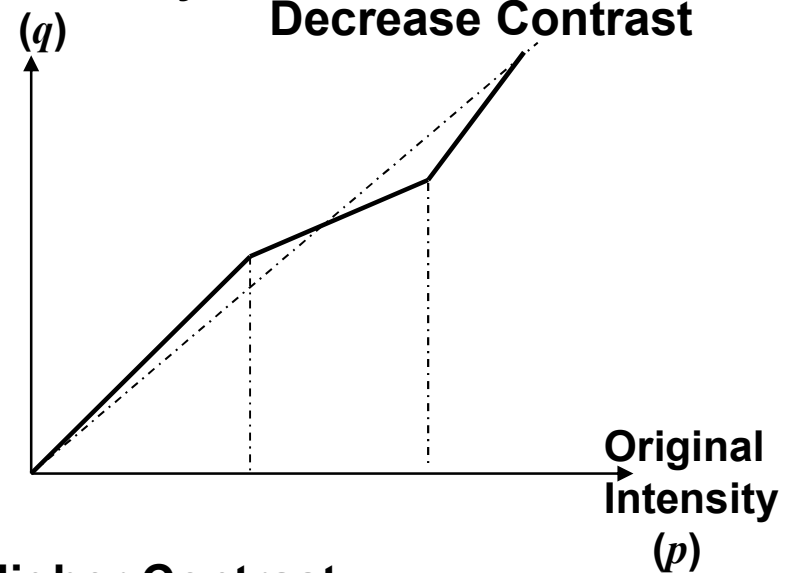
New Intensity



Original



New Intensity



Higher Contrast



# Image Filtering

- Key idea: Images are redundant, a bad pixel can be replaced by a local average
- Examples:
  - Averaging
  - Averaging with limited data validity
  - Averaging using a rotating mask
  - Median filter



# Image Filtering

- Averaging Filter

$$h_1 = \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

4	8	6	11	8
10	5	9	7	10
6	3	6	4	3
9	5	7	9	8
12	2	5	7	4

**5x5 Noisy Image**

	6	7	7	
	7	6	7	
	6	5	6	

**Filtered Image**

# Convolution

- The integral of the product of two functions after one is reversed and shifted
- In mathematical terms for discrete functions

$$f(i, j) = \sum_{(m,n) \in O} g(i-m, j-n)h(m, n)$$

where  $h$  is the convolution mask,  $g$  is the original image and  $O$  defines the size of the mask

- Both noise filtering and edge detection depend on the idea of convolution
- The dimensions of the filter are always odd so that there is always a central pixel

# Convolution

- Example

$$f(i, j) = \sum_{m=1}^{-1} \sum_{n=1}^{-1} g(i-m, j-n) h(m, n)$$

$$h = \frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

$g$

	1	2	3	4	5
1	4	8	6	11	8
2	10	5 $h$	9	7	10
3	6	3	6	4	3
4	9	5	7	9	8
5	12	2	5	7	4

$f$

	1	2	3	4	5
1					
2		6.375	6.875	7.4375	
3		5.6875	6	6.25	
4		5.6875	5.875	6.625	
5					

$$f(2,2) = \frac{1}{16} [4 \times 1 + 10 \times 2 + 6 \times 1 + 8 \times 2 + 5 \times 4 + 3 \times 2 + 6 \times 1 + 9 \times 2 + 6 \times 1] = 6.375$$

$$f(2,3) = \frac{1}{16} [8 \times 1 + 5 \times 2 + 3 \times 1 + 6 \times 2 + 9 \times 4 + 6 \times 2 + 11 \times 1 + 7 \times 2 + 4 \times 1] = 6.875$$

# Image Filtering

- Weighted Averaging Filter

$$h_2 = \frac{1}{10} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

**More weight for central pixel**

$$h_3 = \frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

**More weight for central pixel  
and the 4 neighbors**

- Main Disadvantage of Averaging Filters: Blurring

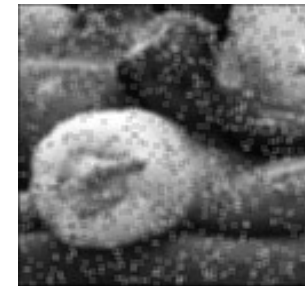
Because of considering pixels in the average that may have different properties than the processed pixel



**Original Image**



**Noisy Image**



**Filtered Image**

# Image Filtering

- Averaging with limited data validity
  - Previous filters were linear
- To avoid blurring, this method is based on the idea that the calculated average should be computed only from points in the neighborhood that satisfy certain condition
- This makes such filter non-linear

$$h(i, j) = \begin{cases} 1 & \text{for } g(m+i, n+j) \in [\text{min}, \text{max}] \\ 0 & \text{otherwise} \end{cases}$$

# Image Filtering

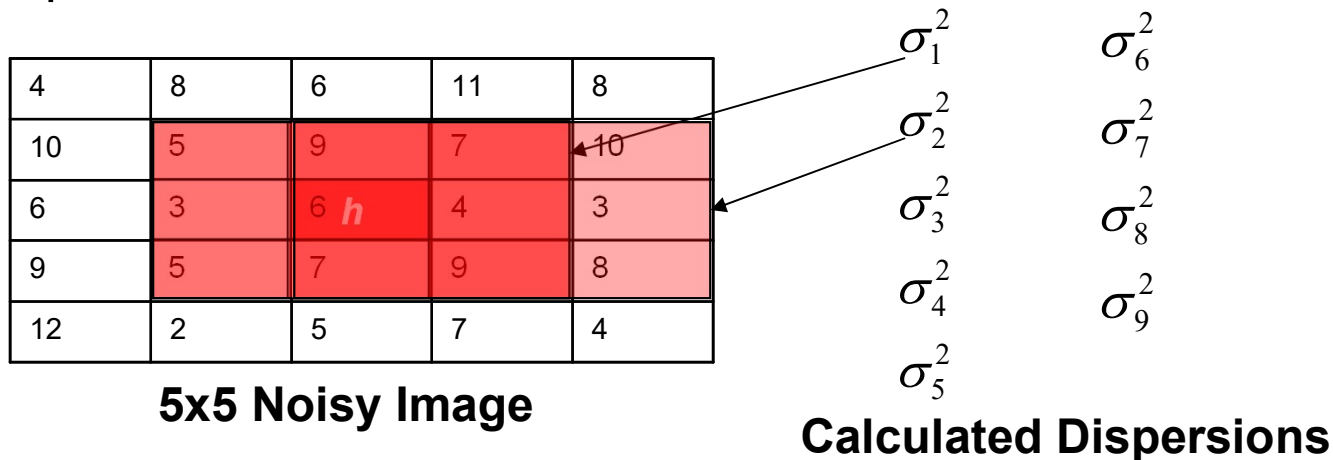
- Averaging using a rotating mask
  - Searches for the homogeneous part of the processed pixel neighborhood to avoid blurring
  - A brightness dispersion  $\sigma^2$  is used as the region homogeneity measure

$$\sigma^2 = \frac{1}{n} \left( \sum_{(i,j) \in R} \left( g(i,j) - \frac{1}{n} \sum_{(i,j) \in R} g(i,j) \right)^2 \right)$$

- To search for the most homogenous part of the neighborhood, the dispersion for all possible mask rotations is calculated
- The position of least dispersion is considered as the best filter for the processed pixel

# Image Filtering

- Example: Consider a 3x3 filter



- Rotating Mask Algorithm
  1. Consider each image pixel  $(i, j)$
  2. Calculate the dispersion for all possible mask rotations around the pixel  $(i, j)$  according to the given equation
  3. Choose the mask with minimum dispersion
  4. Assign to pixel  $(i, j)$  in the output image the average brightness in the chosen mask

# Image Filtering

- Example

**Noisy Image**

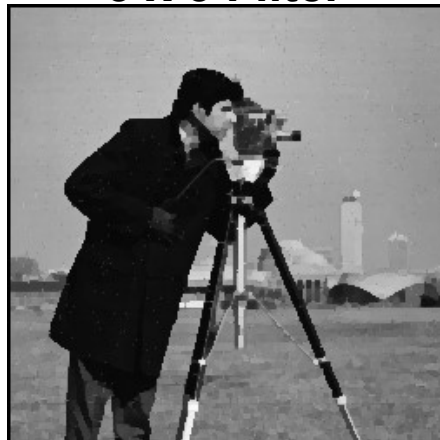


**Filtered using Average Filter (3 x 3)**



**Filtered using Rotating Mask**

**3 x 3 Filter**



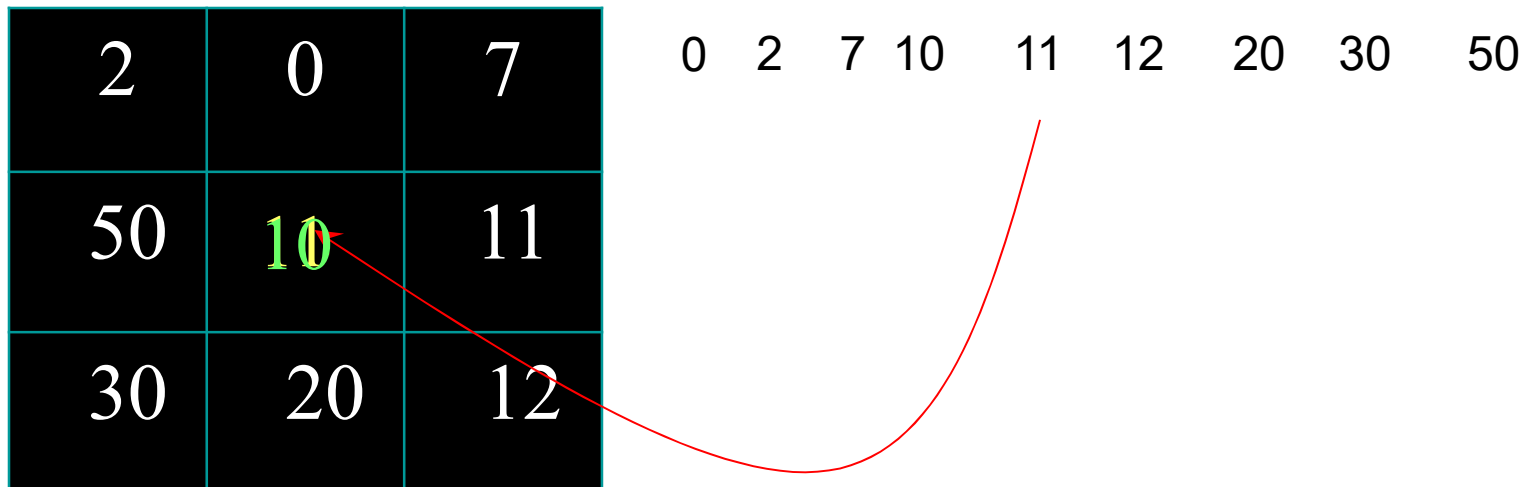
**5 x 5 Filter**





# Image Filtering

- Median Filter
  - In a set of ordered values, the median is the central pixel



- Works very well with salt and pepper noise

# Image Filtering

- Median Filter

**Original Image**



**Noisy Image**



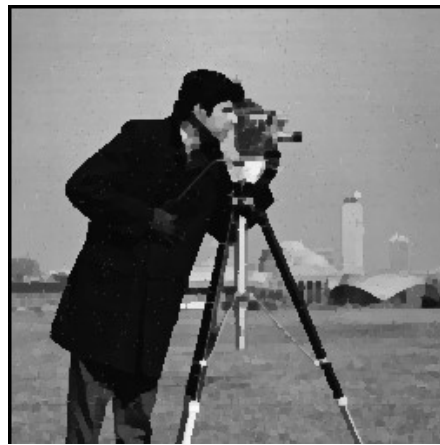
**Filtered Image**



**Noisy Image**



**Filtered using Rotating Mask**



**Filtered using Median Filter**



- Disadvantage: Damages thin lines

# Image Filtering

- Disadvantage of Median Filter: Damages thin lines

0	255	0	0	0
0	255	0	0	0
0	255	0	0	0
0	255	0	0	0
0	255	0	0	0

**Image with a 1-pixel wide  
white line**

0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0

**Filtered using 3x3 median filter**