# Digital Image & Video Processing

Lecture 6

Image Pre-processing
(Image Local Pre-processing: Edge Detection)

Lecturer: Associate Prof. Lý Quốc Ngọc

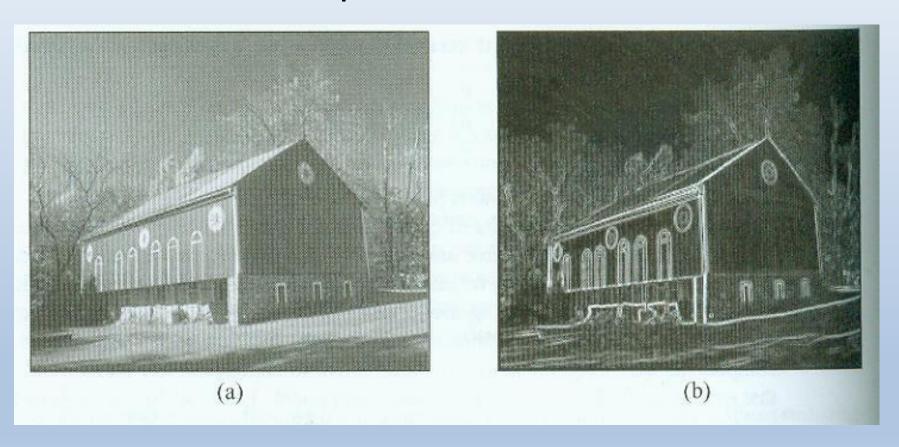
### 3. Image Pre-processing

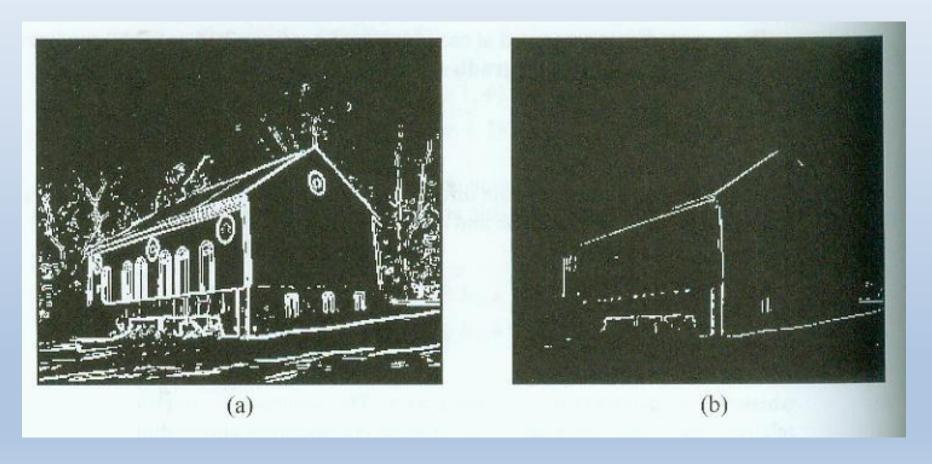
- 3.1. Color Transformations
- 3.2. Geometric Transformations
- 3.3. Local Pre-processing

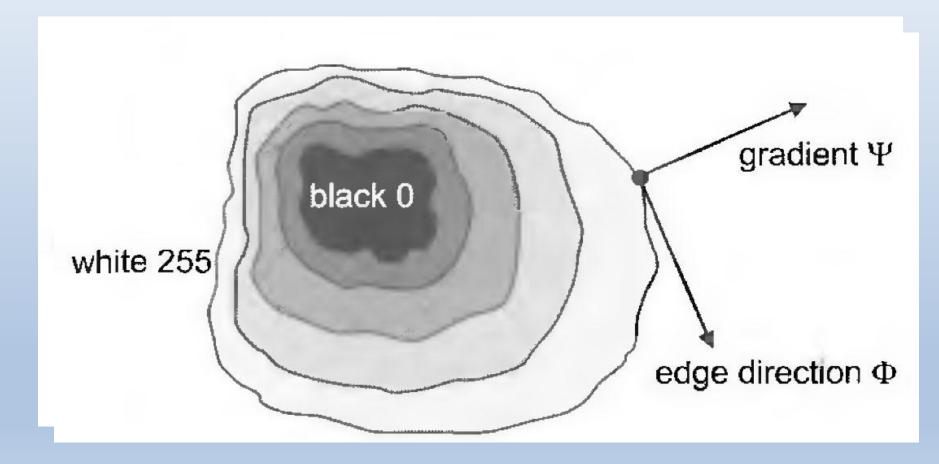
# 3.3. Local Pre-processing

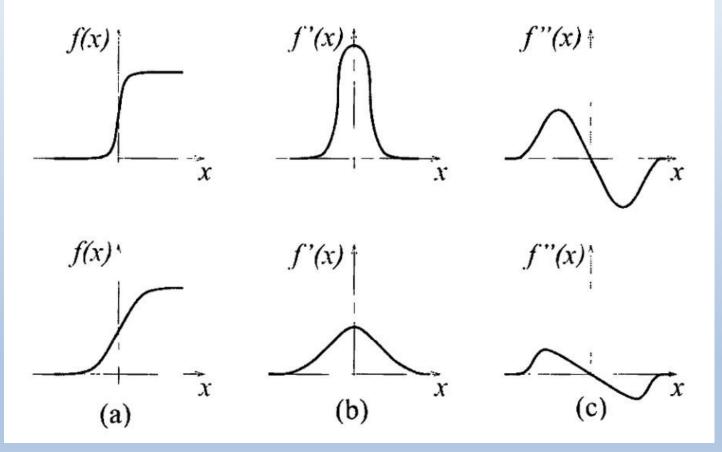
- 3.3.1. Image Smoothing
- 3.3.2. Edge Detection

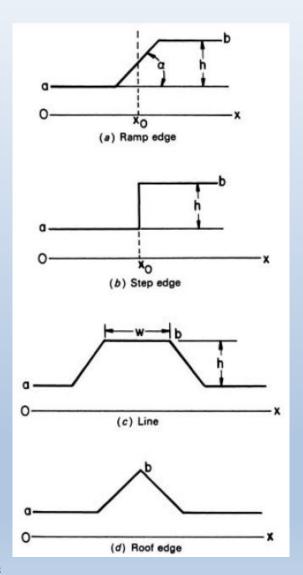
- 3.3.2.1. Gradient operator
- 3.3.1.2. Laplace operator
- 3.3.1.3. Laplace of Gaussian
- 3.3.1.4. Canny method







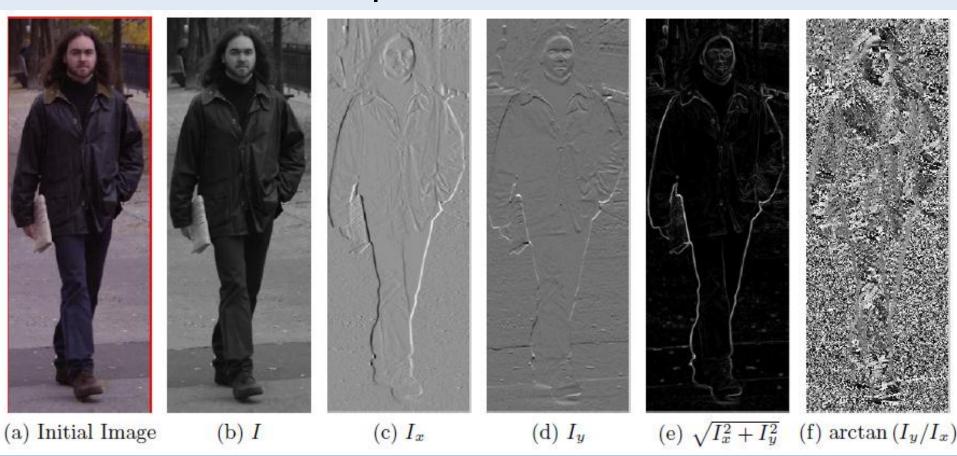


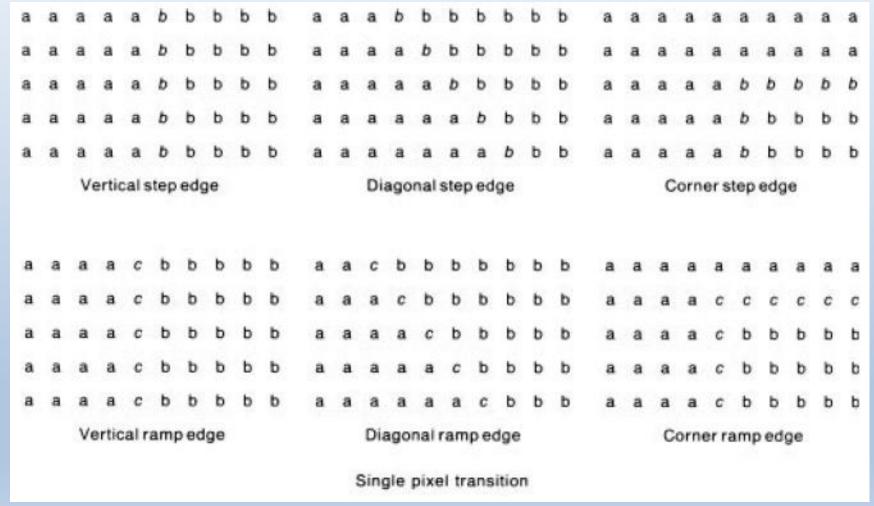


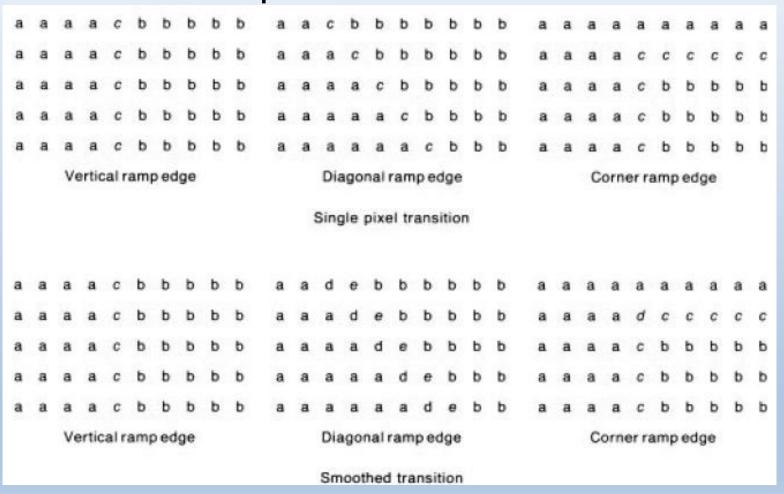
$$(\nabla f)(x, y) = [\partial f / \partial x \ \partial f / \partial y]^T = [f_x \ f_y]^T$$

$$e(x, y) = (f_x^2(x, y) + f_y^2(x, y))^{1/2}$$

$$\phi(x, y) = \arctan(f_x \ f_y)$$







$$c=(a+b)/2,$$

$$d = (3a+b)/4$$

$$e = (a + 3b)/4$$

#### 3.3.2.1. Gradient operator

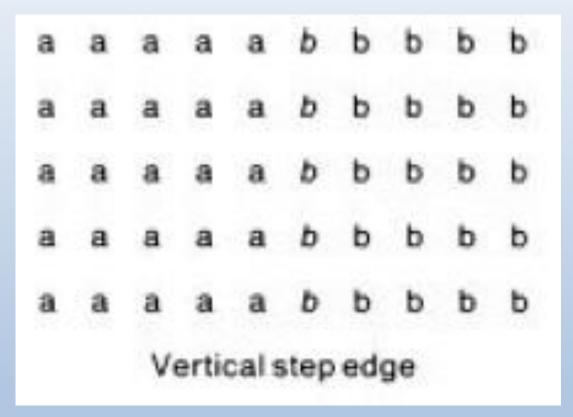
#### **Differencing**

$$f_x(x, y) \approx f(x, y) - f(x+1, y)$$
$$f_y(x, y) \approx f(x, y) - f(x, y-1)$$

#### 3.3.2.1. Gradient operator

#### **Differencing**

h=b-a

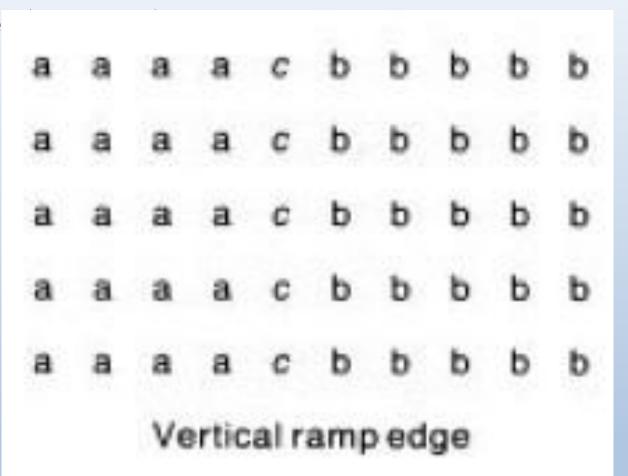


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#### 3.3.2.1. Gradie

### **Differencing**

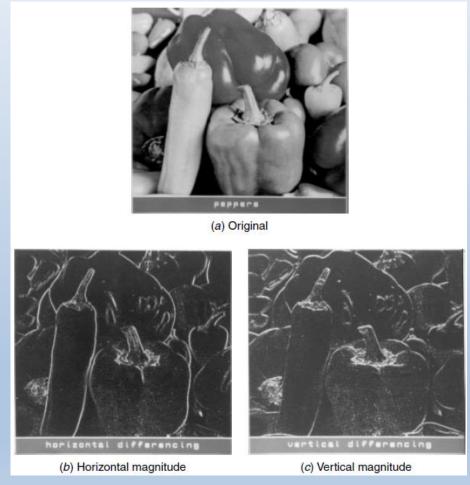
h=b-a



 $0\ 0\ 0\ h/2\ h/2\ 0\ 0$ 

#### 3.3.2.1. Gradient operator

#### Differencing



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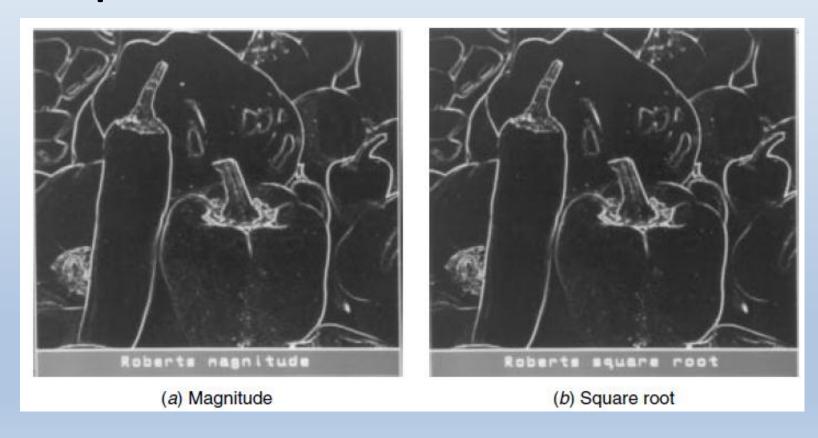
#### 3.3.2.1. Gradient operator

#### **Robert operator**

$$f_x(x, y) \approx f(x, y) - f(x+1, y+1)$$
  
 $f_y(x, y) \approx f(x, y+1) - f(x+1, y)$ 

#### 3.3.2.1. Gradient operator

#### **Robert operator**



#### 3.3.2.1. Gradient operator

#### Differencing (localize edge center of ramp edge)

$$f_x(x, y) \approx f(x-1, y) - f(x+1, y)$$

$$f_{y}(x, y) \approx f(x, y+1) - f(x, y-1)$$

#### 3.3.2.1. Gradient operator

Differencing a a a a a c

(localize edge a a a a a c

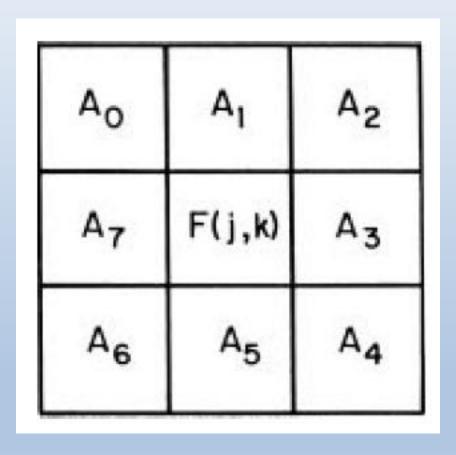
center of a a a a a c

ramp edge) a a a a a c

h=b-a

Vertical ramp edge

 $0 \ 0 \ h/2 \ h \ h/2 \ 0 \ 0$ 



#### 3.3.2.1. Gradient operator

#### Prewitt operator (k=1)

$$f_x \approx \frac{1}{k+2} [(A_2 + kA_3 + A_4) - (A_0 + kA_7 + A_6)]$$

$$f_y \approx \frac{1}{k+2} [(A_0 + kA_1 + A_2) - (A_6 + kA_5 + A_4)]$$

#### 3.3.2.1. Gradient operator

#### Sobel operator (k=2)

$$f_x \approx \frac{1}{k+2} [(A_2 + kA_3 + A_4) - (A_0 + kA_7 + A_6)]$$

$$f_y \approx \frac{1}{k+2} [(A_0 + kA_1 + A_2) - (A_6 + kA_5 + A_4)]$$

#### 3.3.2.1. Gradient operator

Frei-Chen  $(k=2^{1/2})$ 

$$f_x \approx \frac{1}{k+2} [(A_2 + kA_3 + A_4) - (A_0 + kA_7 + A_6)]$$

$$f_y \approx \frac{1}{k+2} [(A_0 + kA_1 + A_2) - (A_6 + kA_5 + A_4)]$$

#### 3.3.2.1. Gradient operator

**Prewitt** Sobel Frei-Chen h=b-a Vertical ramp edge

 $0 \ 0 \ h/2 \ h \ h/2 \ 0 \ 0$ 

### 3.3.2.1. Gradient operator

**Prewitt** 

Sobel

Frei-Chen

h=b-a

$$0 \frac{h}{\sqrt{2}(2+k)} \frac{h}{\sqrt{2}} \frac{\sqrt{2}(1+k)h}{(2+k)} \frac{h}{\sqrt{2}} \frac{h}{\sqrt{2}(2+k)} 0$$

Frei - Chen : 
$$k = \sqrt{2} \Rightarrow \frac{\sqrt{2}(1+k)h}{(2+k)} = h$$

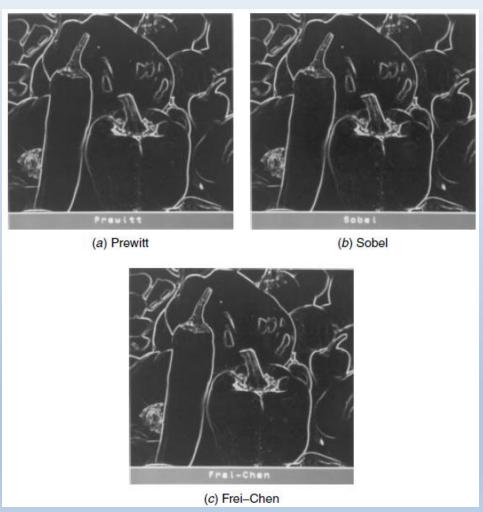
Prewitt: 
$$k = 1 \Rightarrow \frac{\sqrt{2}(2)h}{(3)} = 0.94 \times h$$

Sobel: 
$$k = 2 \Rightarrow \frac{\sqrt{2(3)h}}{(4)} = 1.06 \times h$$

$$h = b - a$$

### 3.3.2.1. Gradient operator

**Prewitt-Sobel-FreiChen** 



### 3.3.2.1. Gradient operator

Convolution operator

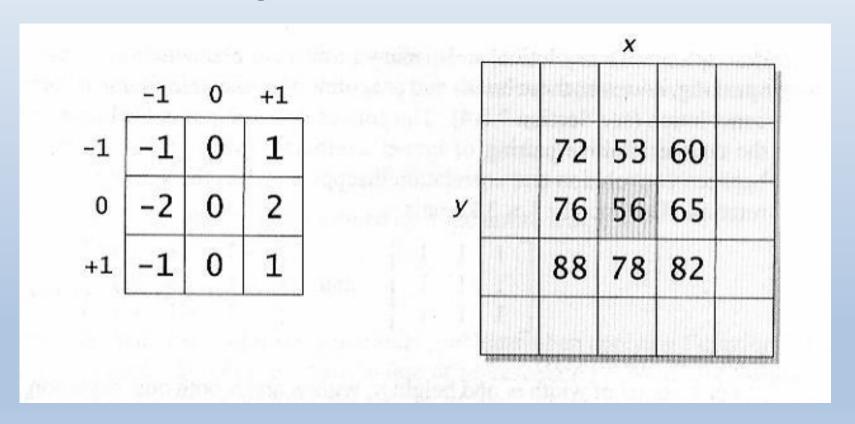
$$g = f * h$$

$$g(x, y) = \sum_{i} \sum_{j} f(x - i, y - j).h(i, j),$$

$$(i, j) \in O$$

#### 3.3.2.1. Gradient operator

#### Comsider image f and filter h as follows:



#### 3.3.2.1. Gradient operator

Ouput image g computed at (x,y):

$$g(x, y) = h(-1,-1)f(x+1, y+1) + h(0,-1)f(x, y+1) +$$

$$h(1,-1)f(x-1, y+1) + h(-1,0)f(x+1, y) +$$

$$h(0,0)f(x, y) + h(1,0)f(x-1, y) +$$

$$h(-1,1)f(x+1, y-1) + h(0,1)f(x, y-1) +$$

$$h(1,1)f(x-1, y-1)$$

$$g(x, y) = 40$$

#### 3.3.2.1. Gradient operator

#### Pixel difference

$$f_{x} \approx \hat{f}_{x} = f * W_{x}$$

$$f_{y} \approx \hat{f}_{y} = f * W_{y}$$

$$W_{x} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix}, W_{y} = \begin{bmatrix} 0 & -1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

#### 3.3.2.1. Gradient operator

Separated Pixel difference

$$f_{x} \approx \hat{f}_{x} = f * W_{x}$$

$$f_{y} \approx \hat{f}_{y} = f * W_{y}$$

$$W_{x} = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix}, W_{y} = \begin{bmatrix} 0 & -1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

#### 3.3.2.1. Gradient operator

### Robert operator

$$f_{x} \approx \hat{f}_{x} = f * W_{x}$$

$$f_{y} \approx \hat{f}_{y} = f * W_{y}$$

$$W_{x} = \begin{bmatrix} 0 & 0 & -1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, W_{y} = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

#### 3.3.2.1. Gradient operator

#### Prewitt operator

$$f_{x} \approx \hat{f}_{x} = f * W_{x}$$

$$f_{y} \approx \hat{f}_{y} = f * W_{y}$$

$$W_{x} = \frac{1}{3} \begin{bmatrix} 1 & 0 & -1 \\ 1 & 0 & -1 \\ 1 & 0 & -1 \end{bmatrix}, W_{y} = \frac{1}{3} \begin{bmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

#### 3.3.2.1. Gradient operator

#### Sobel operator

$$f_{x} \approx \hat{f}_{x} = f * W_{x}$$

$$f_{y} \approx \hat{f}_{y} = f * W_{y}$$

$$W_{x} = \frac{1}{4} \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix}, W_{y} = \frac{1}{4} \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

#### 3.3.2.1. Gradient operator

Frei-chen operator

$$f_x \approx \hat{f}_x = f * W_x, f_y \approx \hat{f}_y = f * W_y,$$

$$W_{x} = \frac{1}{2 + \sqrt{2}} \begin{bmatrix} 1 & 0 & -1 \\ \sqrt{2} & 0 & -\sqrt{2} \\ 1 & 0 & -1 \end{bmatrix},$$

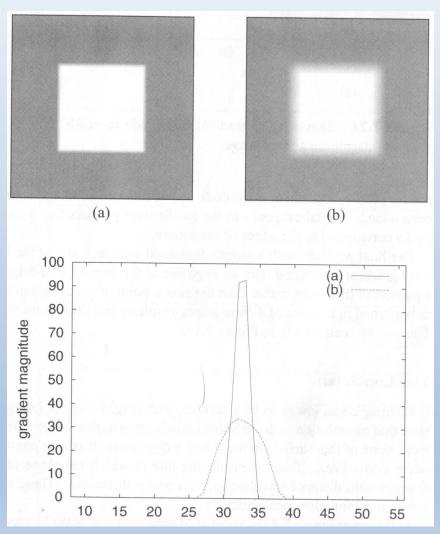
$$W_{y} = \frac{1}{2 + \sqrt{2}} \begin{bmatrix} -1 & -\sqrt{2} & -1 \\ 0 & 0 & 0 \\ 1 & \sqrt{2} & 1 \end{bmatrix}$$

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$$(\nabla^2 f)(x, y) = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$
  

$$(\nabla^2 f)(x, y) \approx f(x+1, y) + f(x-1, y) +$$
  

$$f(x, y+1) + f(x, y-1) - 4f(x, y)$$



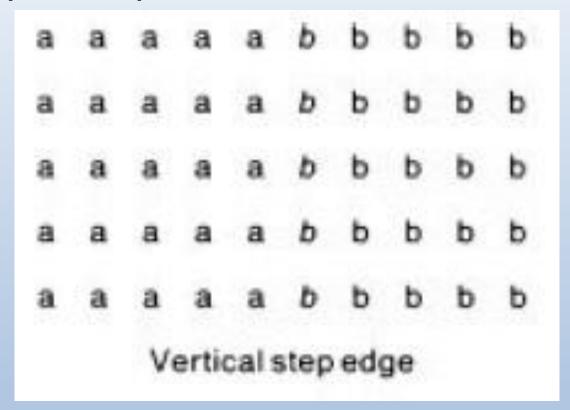
$$\nabla^2 f \approx f * Laplace$$

$$Laplace = \begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\nabla^{2} f \approx f * Laplace$$

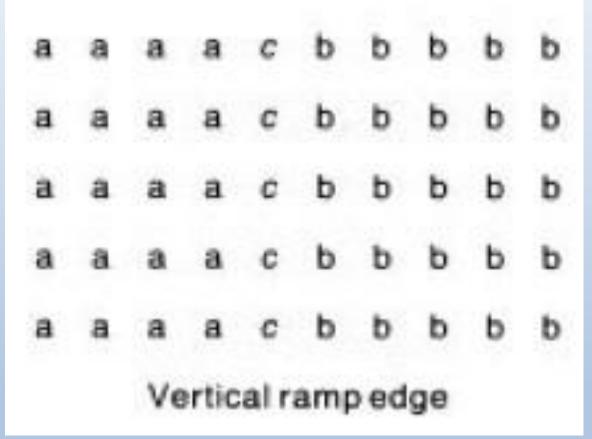
$$Laplace = \begin{bmatrix} 1 & 1 & 1 \\ 1 & -8 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

#### 3.3.2.2. Laplace operator



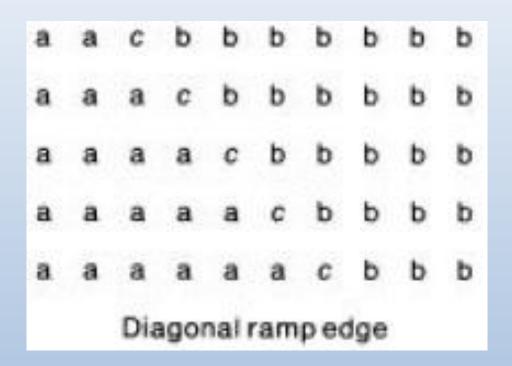
0 - 3h/8 3h/8 0

3.3.2.2. Laplace operator



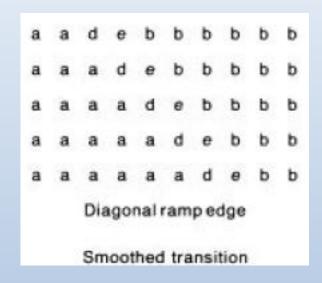
0 - 3h/160 3h/160

#### 3.3.2.2. Laplace operator

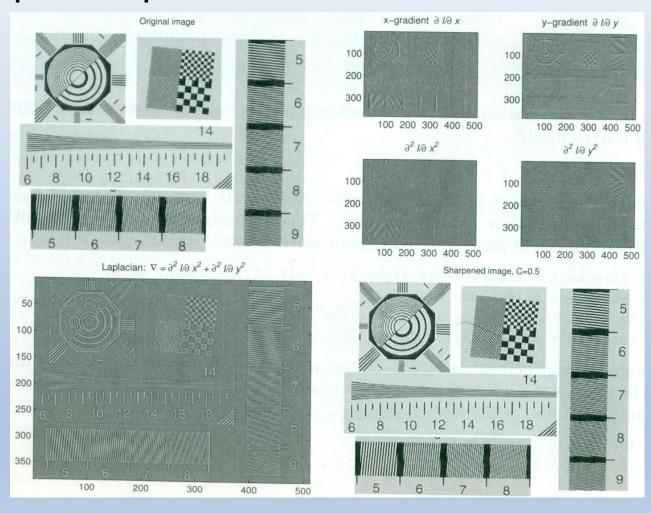


 $0 - h/8 - h/8 \ 0 \ h/8 \ h/8 \ 0$ 

#### 3.3.2.2. Laplace operator



 $0 - h/16 - h/8 - h/16 \ 0 \ h/16 \ h/8 \ h/16 \ 0$ 



#### 3.3.2.2. Laplace of Gaussian operator

$$\nabla^2 [G(x, y, \sigma) * f(x, y)] = [\nabla^2 G(x, y, \sigma)] * f(x, y)$$

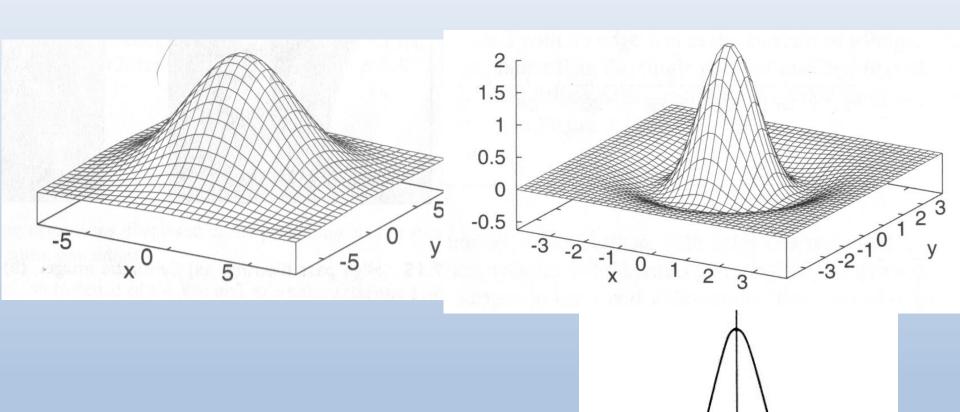
$$G(x, y, \sigma) = \frac{1}{2\pi\sigma^2} e^{-(x^2 + y^2)/2\sigma^2}$$

$$\nabla^2 G(x, y, \sigma) = \frac{\partial^2 G}{\partial x^2} + \frac{\partial^2 G}{\partial y^2}$$

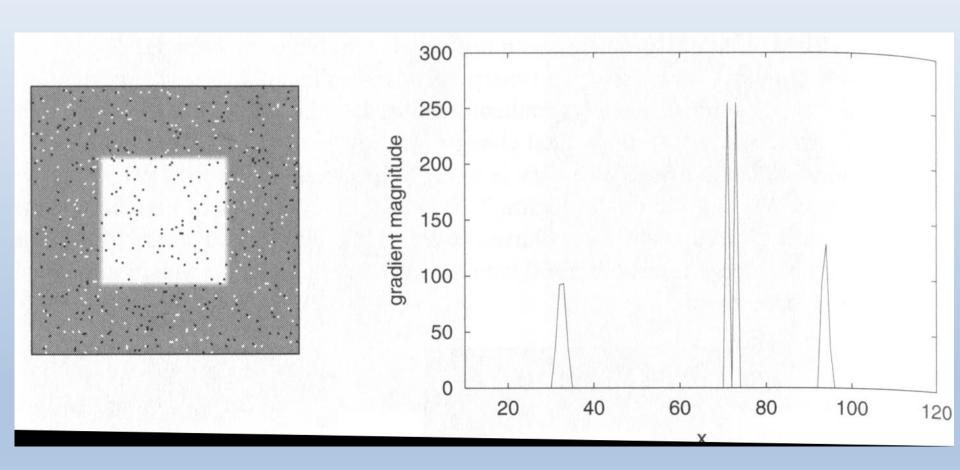
$$= \frac{1}{\pi \sigma^4} \left[ \frac{x^2 + y^2}{2\sigma^2} - 1 \right] e^{-\frac{x^2 + y^2}{2\sigma^2}}$$

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#### 3.3.2.2. Laplace of Gaussian operator



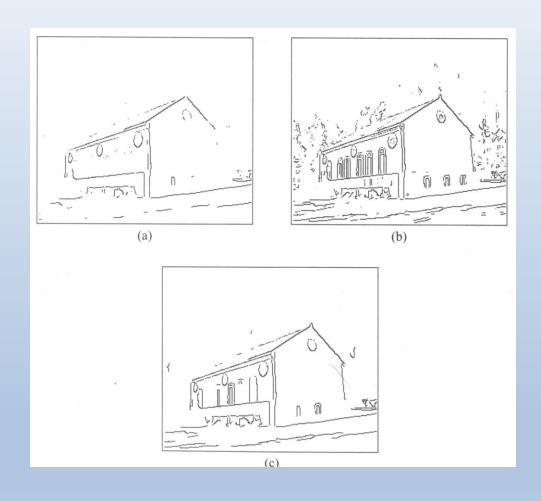
#### 3.3.2.2. Laplace of Gaussian operator



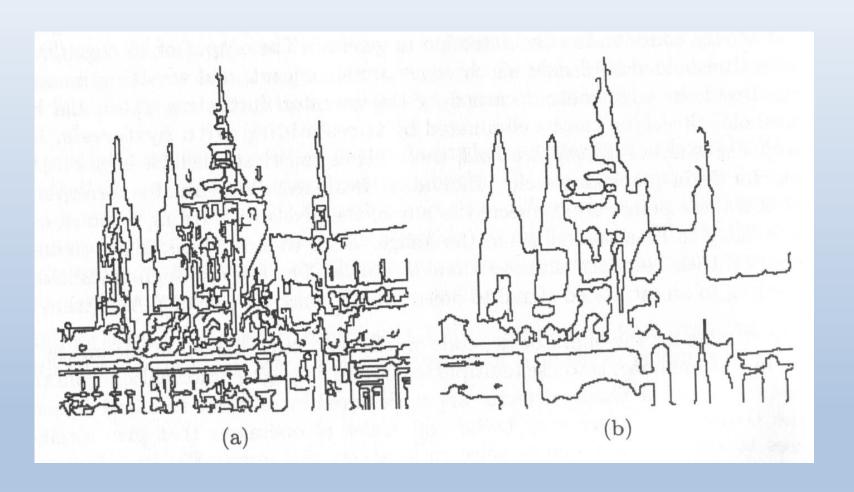
#### 3.3.2.2. Canny edge detection method

- **1.**  $G(x, y, \sigma) * f(x, y)$
- **2.**  $n = \nabla (G * f)$
- 3. Find the location of the edges using non-maximal suppression approach.
- **4.** Compute the magnitude of the edge is measured as  $|\nabla(G*f)|$
- 5. Threshold edges in the image with hysteresis to eliminate spurious responses.
- **6.** Repeat steps 1-5 for ascending values of the standard deviation  $\sigma$ .
- 7. Aggregate the final information about edges at multiple scale using 'feature systhesis' approach

#### 3.3.2.2. Canny edge detection method



#### 3.3.2.2. Canny edge detection method

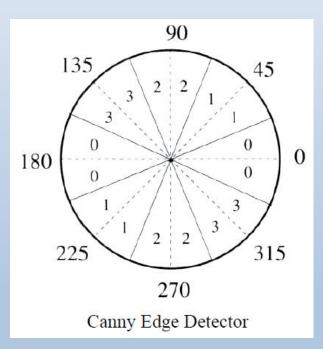


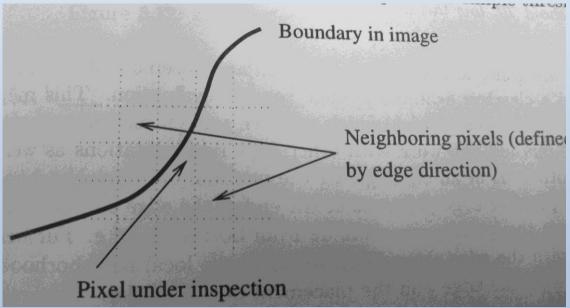
#### 3.3.2.2. Canny edge detection method

#### Non-maximal suppression of directional edge data

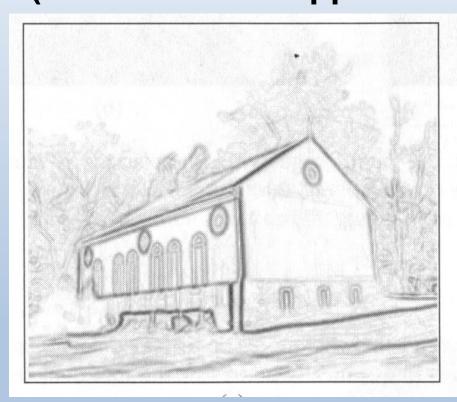
- **3.** Find the location of the edges using non-maximal suppression approach.
- **3.1**. For each pixel with non-zero edge magnitude, inspect two adjacent pixels indicated the direction of its edge.
- **3.2**. If the edge magnitude of either of these two exceeds that of the pixel under inspection, mark it for deletion.
- 3.3. When all pixels have been inspected, re-scan the image and erase to zero all edge data marked for deletion.

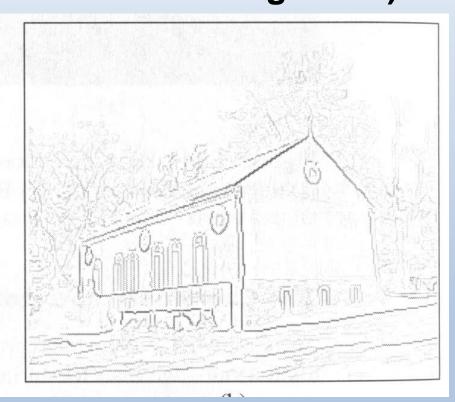
# 3.3.2.2. Canny edge detection method (Non-maximal suppression of directional edge data)





# 3.3.2.2. Canny edge detection method (Non-maximal suppression of directional edge data)





#### 3.3.2.2. Canny edge detection method

#### Hysteresis to filter output of an edge detector

- **5.** Threshold edges in the image with hysteresis to eliminate spurious responses.
- **5.1**. Mark all edges with magnitude greater than  $t_1$ .
- **5.2**. Scall all pixels with edge magnitude in the range [t<sub>0</sub> t<sub>1</sub>].
- **5.3**. If such a pixel borders another marked as an edge, the mark it too.
- **5.4**. Repeat from step 5.2 until stability.

# 3.3.2.2. Canny edge detection method Hysteresis to filter output of an edge detector

