

# Lecture 07 – Spatial filtering II

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## Agenda

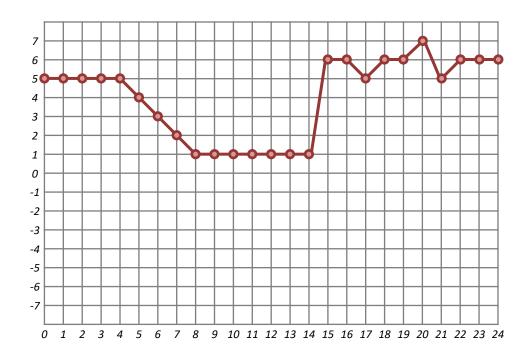


- Derivatives of 1D discrete functions
- The Laplacian
- Laplacian variations
- The Gradient
- Roberts cross-gradient operators
- Prewitt and Sobel operators



## **DERIVATIVES OF 1D DISCRETE FUNCTIONS**





First order derivative of a 1D function f(x):

$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

Second order derivative of a 1D function f(x):

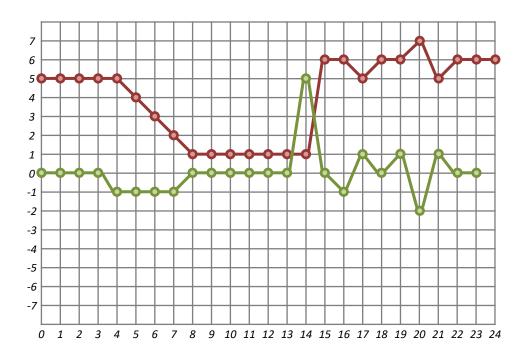
$$\frac{\partial^2 f}{\partial x^2} = f(x+1) + f(x-1) - 2f(x)$$

5 5 5 5 5 4 3 2 1 1 1 1 1 1 1 6 6 5 6 6 7 5 6 6 6 Signal

First order derivative

Second order derivative





First order derivative of a 1D function f(x):



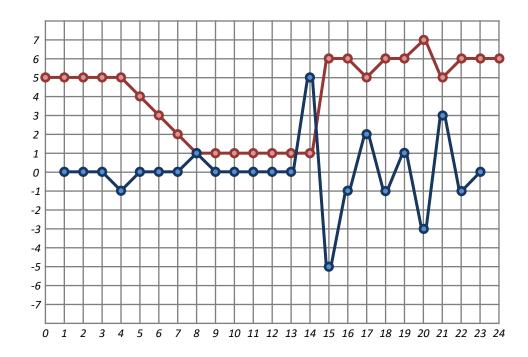
$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

Second order derivative of a 1D function f(x):

$$\frac{\partial^2 f}{\partial x^2} = f(x+1) + f(x-1) - 2f(x)$$

Signal
First order derivative
Second order derivative





First order derivative of a 1D function f(x):

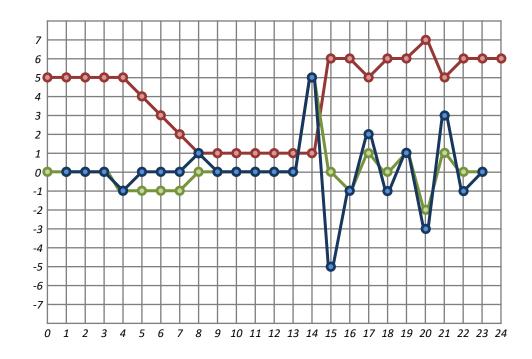
$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

Second order derivative of a 1D function f(x):

$$\frac{\partial^2 f}{\partial x^2} = f(x+1) + f(x-1) - 2f(x)$$

Signal
First order derivative
Second order derivative





First order derivative of a 1D function f(x):



$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

Second order derivative of a 1D function f(x):

$$\frac{\partial^2 f}{\partial x^2} = f(x+1) + f(x-1) - 2f(x)$$

Signal
First order derivative
Second order derivative



# THE LAPLACIAN

## The Laplacian



• The Laplacian of a two-dimensional function f(x, y) is:

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

• If we separate the Laplacian into the x and y directions, we have:

$$\frac{\partial^2 f}{\partial x^2} = f(x+1,y) + f(x-1,y) - 2f(x,y)$$
$$\frac{\partial^2 f}{\partial y^2} = f(x,y+1) + f(x,y-1) - 2f(x,y)$$

Thus, the discrete Laplacian of two variables is:

$$\nabla^2 f = f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1) - 4f(x,y)$$

	-1	0	1
-1	0	1	0
0	1	-4	1
1	0	1	0

# Laplacian variations



	-1	U	1
-1	0	1	0
0	1	-4	1

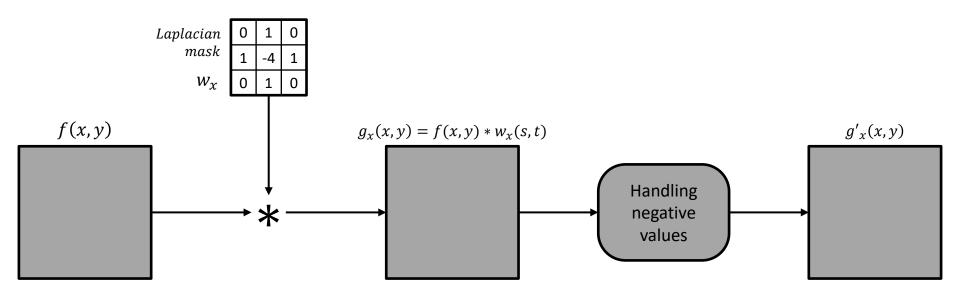
			_
-1	0	-1	0
0	-1	4	-1
1	0	-1	0

	-1	0	1
-1	1	1	1
0	1	-8	1
1	1	1	1

	-1	0	1
-1	-1	-1	-1
0	-1	8	-1
1	-1	-1	-1

# The Laplacian – how to apply







# THE GRADIENT

### The Gradient



• The gradient of a two-dimensional function f(x, y) is:

$$\nabla f \equiv \begin{bmatrix} g_x \\ g_y \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix},$$

 $\frac{\partial f}{\partial x} = f(x, y) - f(x+1, y), \quad \frac{\partial f}{\partial y} = f(x, y) - f(x, y+1)$   $1 \quad \boxed{1} \quad \boxed{0}$ 

• The magnitude (size) of the gradient vector  $(\nabla f)$ , M(x, y) is:

$$M(x,y) = mag(\nabla f) = \sqrt{g_x^2 + g_y^2}$$

Or it can be approximated by absolute values:

$$M(x,y) \approx |g_x| + |g_y|$$

	0	1
0	1	-1
1	0	0

# The Gradient – Roberts cross-gradient operators



Roberts cross-gradient operators consider diagonal differences:

$$\frac{\partial f}{\partial x} = f(x, y) - f(x + 1, y + 1),$$

$$\frac{\partial f}{\partial x} = f(x,y) - f(x+1,y+1), \qquad \frac{\partial f}{\partial y} = f(x+1,y) - f(x,y+1)$$

# The Gradient – Prewitt and Sobel operators



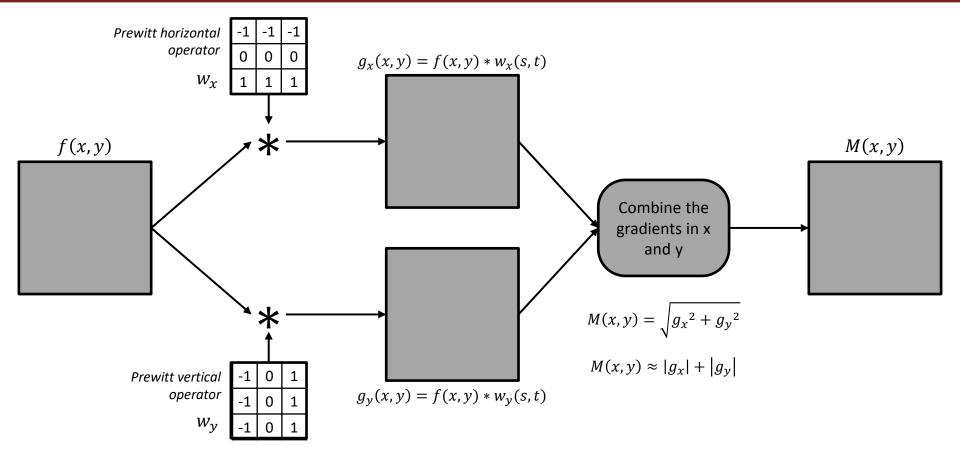
Prewitt:

Sobel:

$$g_x$$
 -1 0 1  
-1 -1 -2 -1  
0 0 0 0  
1 1 2 1

## The Gradient – how to apply





## Bibliography



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