### Digital Image Processing

Lecture # 3D Spatial Filtering

#### Contents

- Sharpening Spatial Filters
- Image Enhancement using
  - 2<sup>nd</sup> Derivative
  - 1<sup>st</sup> Derivative
- Combining Spatial Enhancement Methods

### Sharpening Spatial Filters

Previously we have looked at smoothing filters which remove fine detail

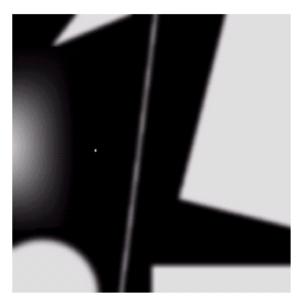
Sharpening spatial filters seek to highlight fine detail

- Remove blurring from images
- Highlight edges

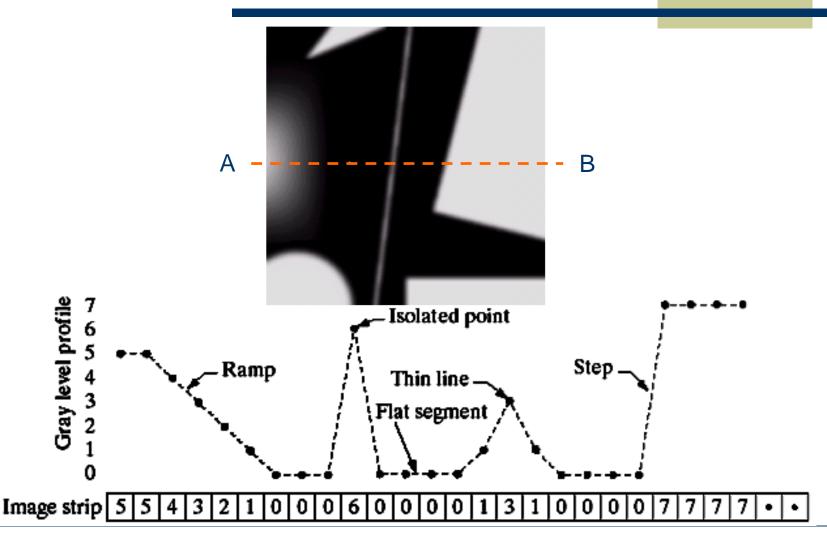
Sharpening filters are based on spatial differentiation

### Spatial Differentiation

 Let's consider a simple 1 dimensional example



### Spatial Differentiation



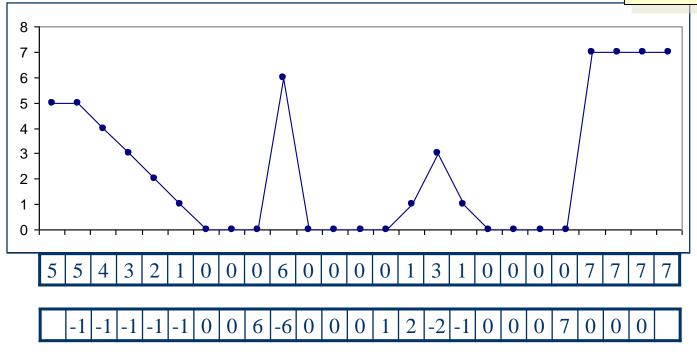
#### 1st Derivative

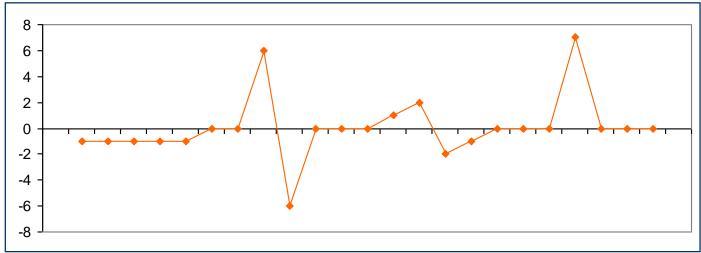
The 1<sup>st</sup> derivative of a function is given by:

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

Its just the difference between subsequent values and measures the rate of change of the function

#### 1<sup>st</sup> Derivative





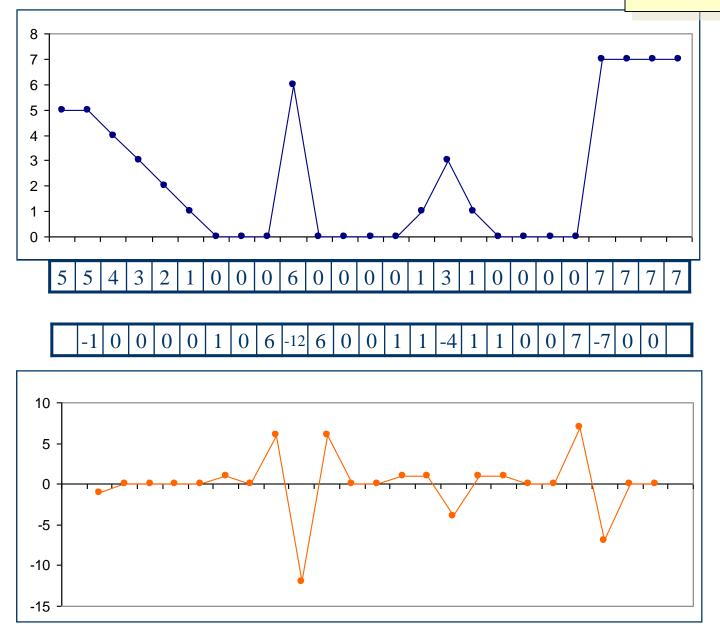
#### 2<sup>nd</sup> Derivative

The 2nd derivative of a function is given by:

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

Simply takes into account the values both before and after the current value

#### 2<sup>nd</sup> Derivative



#### 2<sup>nd</sup> Derivative for Image Enhancement

The 2nd derivative is more useful for image enhancement than the 1st derivative - Stronger response to fine detail

We will come back to the 1st order derivative later on

The first sharpening filter we will look at is the Laplacian

#### The Laplacian is defined as follows:

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

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

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

#### So, the Laplacian can be given as follows:

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

Can we implement it using a filter/ mask?

0	1	0
1	-4	1
0	1	0

0	1	0	1	1	1
1	-4	1	1	-8	1
0	1	0	1	1	1
0	-1	0	-1	-1	-1
-1	4	-1	-1	8	-1
0	-1	0	-1	-1	-1

a b c d

#### FIGURE 3.39

(a) Filter mask used to implement the digital Laplacian, as defined in Eq. (3.7-4). (b) Mask used to implement an extension of this equation that includes the diagonal neighbors. (c) and (d) Two other implementations of the Laplacian.

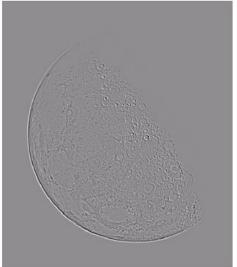
Applying the Laplacian to an image we get a new image that highlights edges and other discontinuities



Original Image



Laplacian Filtered Image

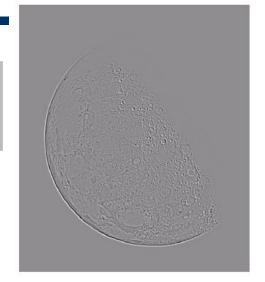


Laplacian
Filtered Image
Scaled for Display

# Laplacian Image Enhancement

The result of a Laplacian filtering is not an enhanced image

To generate the final enhanced image



Laplacian
Filtered Image
Scaled for Display

$$g(x, y) = \frac{f(x, y) - \nabla^2 f, w_5 < 0}{f(x, y) + \nabla^2 f, w_5 > 0}$$

### Laplacian Image Enhancement



In the final sharpened image edges and fine detail are much more obvious

# Laplacian Image Enhancement





# Simplified Image Enhancement

 The entire enhancement can be combined into a single filtering operation

$$g(x, y) = f(x, y) - \nabla^{2} f$$

$$= f(x, y) - [f(x+1, y) + f(x-1, y) + f(x, y+1) + f(x, y+1)$$

# Simplified Image Enhancement

 The entire enhancement can be combined into a single filtering operation

$$g(x, y) = f(x, y) - \nabla^{2} f$$

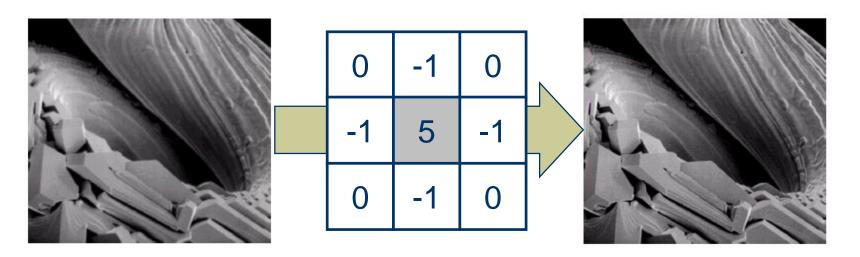
$$= 5 f(x, y) - f(x+1, y) - f(x-1, y)$$

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

0	-1	0
-1	5	-1
0	-1	0

# Simplified Image Enhancement

 This gives us a new filter which does the whole job for us in one step



# Use of first derivatives for image enhancement: The Gradient

• The **gradient** of a function f(x,y) is defined as

$$\nabla \mathbf{f} = \begin{bmatrix} G_x \\ G_y \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

# Use of first derivatives for image enhancement: The Gradient

The magnitude of this vector is given by:

$$\nabla f = mag(\nabla f)$$

$$= \left[G_x^2 + G_y^2\right]^{1/2}$$

$$= \left[\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2\right]^{1/2}$$

For practical reasons this can be simplified as:

$$\nabla f \approx \left| G_{x} \right| + \left| G_{y} \right|$$

### **Gradient Operators**

There is some debate as to how best to calculate these gradients

#### Simplest Operator

$$\frac{\partial f}{\partial y} = (z_8 - z_5), \frac{\partial f}{\partial x} = (z_6 - z_5)$$

$$\nabla f = \sqrt{(z_8 - z_5)^2 + (z_6 - z_5)^2}$$

$$\nabla f \approx |(z_8 - z_5)| + |(z_6 - z_5)|$$

$z_1$	$z_2$	Z <sub>3</sub>
z <sub>4</sub>	$z_5$	z <sub>6</sub>
z <sub>7</sub>	$z_8$	Z9

#### **Gradient Operators**

#### Prewitt Operator

$$\nabla f \approx \left| (z_7 + z_8 + z_9) - (z_1 + z_2 + z_3) \right| + \left| (z_3 + z_6 + z_9) - (z_1 + z_4 + z_7) \right|$$

$z_1$	$z_2$	<i>z</i> <sub>3</sub>
Z <sub>4</sub>	Z <sub>5</sub>	z <sub>6</sub>
z <sub>7</sub>	$z_8$	Z <sub>9</sub>

$$\frac{\partial f}{\partial y} = \begin{array}{|c|c|c|c|c|} \hline -1 & -1 & -1 \\ \hline 0 & 0 & 0 \\ \hline 1 & 1 & 1 \end{array}$$

	-1	0	1
· =	-1	0	1
	-1	0	1

Extract vertical edges

### **Gradient Operators**

#### Sobel Operator

$$\frac{\partial f}{\partial y} = \begin{array}{|c|c|c|c|c|} \hline -1 & -2 & -1 \\ \hline 0 & 0 & 0 \\ \hline 1 & 2 & 1 \\ \hline \end{array}$$

$$\frac{\partial f}{\partial x} =$$

-1	0	1
-2	0	2
-1	0	1

Extract horizontal edges

Extract vertical edges

Emphasize more the current point (x direction)

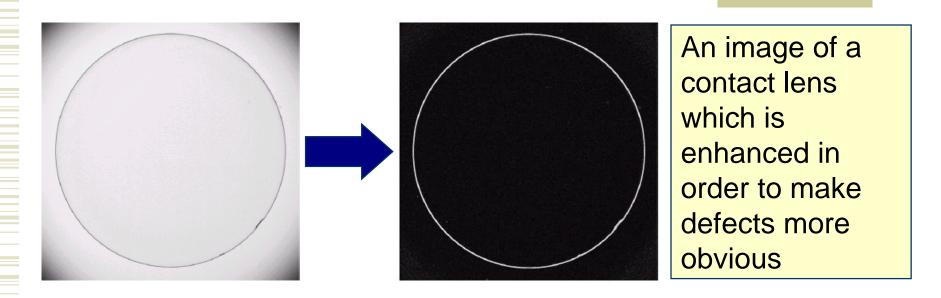
$$\nabla f \approx \left| (z_7 + 2z_8 + z_9) - (z_1 + 2z_2 + z_3) \right| + \left| (z_3 + 2z_6 + z_9) - (z_1 + 2z_4 + z_7) \right|$$

Emphasize more the current point (y direction)

$z_1$	$z_2$	Z <sub>3</sub>
Z <sub>4</sub>	Z <sub>5</sub>	z <sub>6</sub>
z <sub>7</sub>	$z_8$	Z9

Pixel Arrangement

### Sobel Operator: Example

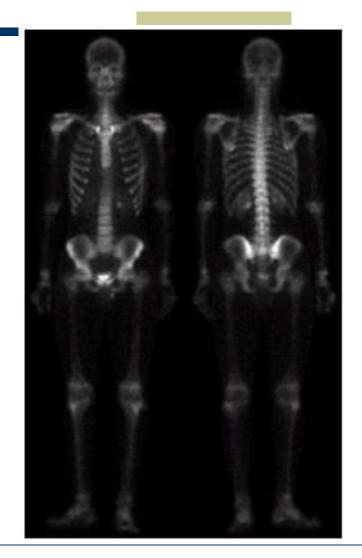


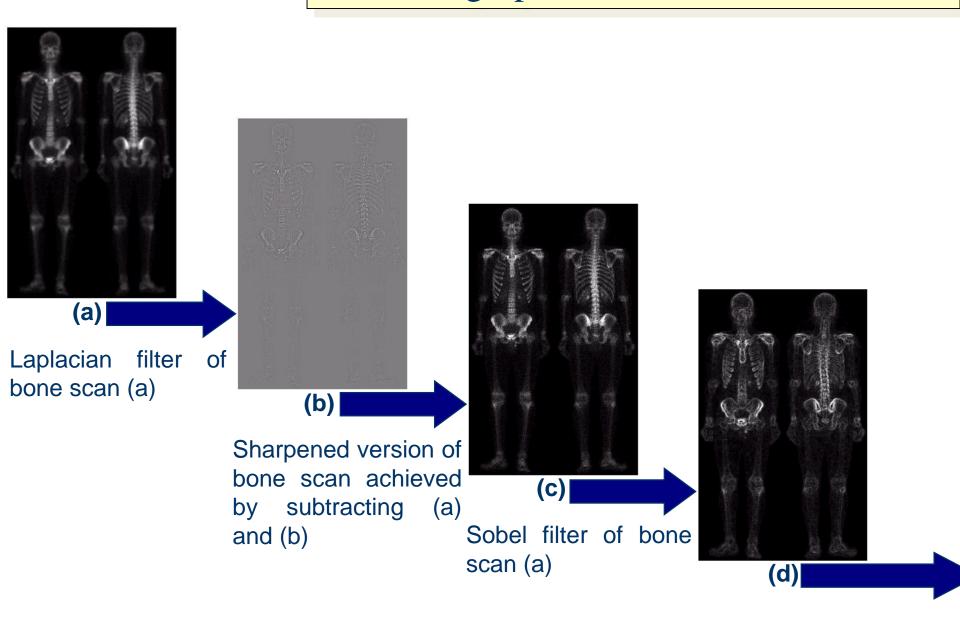
Sobel filters are typically used for edge detection

Successful image enhancement is typically not achieved using a single operation

Rather we combine a range of techniques in order to achieve a final result

This example will focus on enhancing the bone scan





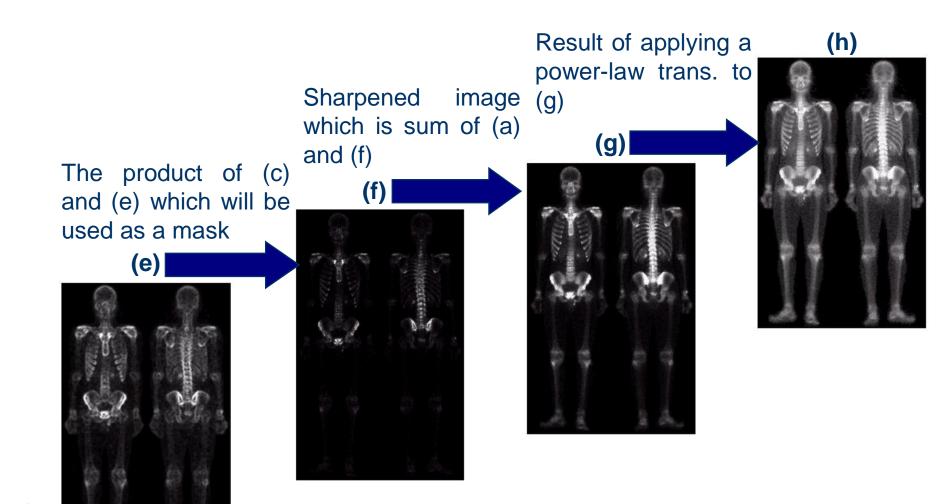
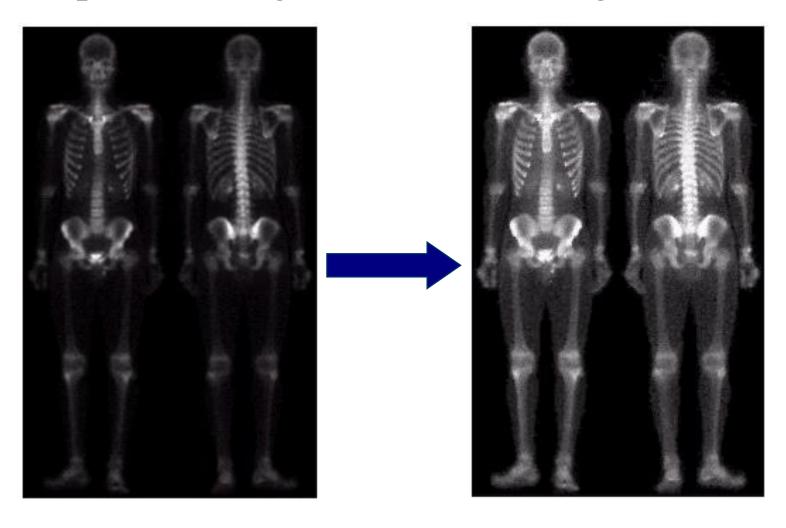


Image (d) smoothed with a 5\*5 averaging filter

#### Compare the original and final images



### Acknowledgements

- Digital Image Processing", Rafael C. Gonzalez & Richard E. Woods, Addison-Wesley, 2002
- Peters, Richard Alan, II, Lectures on Image Processing, Vanderbilt University, Nashville, TN,
   April 2008
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- Computer Vision for Computer Graphics, Mark Borg