

PAM3012
 Digital Image Processing for Radiographers
 Image Enhancement in the Spatial Domain
 (Part III)
 Spatial Filtering

In this lecture

- ★Recap: Spatial Enhancement
- ★ Spatial Filtering
- ★ Smoothing Filters
- ★ Sharpening Filters

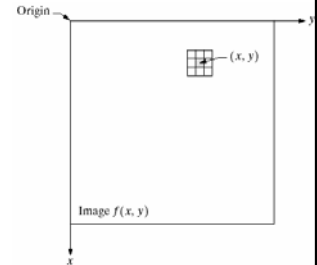
Background

- Procedures that operate directly on the aggregate of pixels composing an image

$$g(x,y) = T[f(x,y)]$$

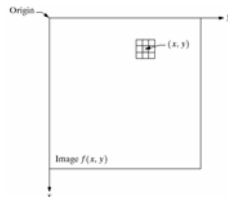
Background

- A neighborhood about (x,y) is defined by using a square subimage area centered at (x,y) .
- Operator T is applied to each location



Background

- If neighborhood is greater than 1 X 1
- Values of f in a predefined neighborhood of (x,y) determine the value of g at (x,y) .
- Values of surrounding pixels (mask) determine nature of process in each pixel
- Called *Mask processing* or *filtering*



Spatial Filtering

Modifying brightness of pixels depending on brightness of surrounding pixels & predefined sub-image

Definitions

Pixel values 'beneath' sub-image

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

- Sub-image size $m \times n$
- Values in sub-image are called coefficients

Definitions

Often simplified

Pixel values 'beneath' sub-image

z_1	z_2	z_3
z_4	z_5	z_6
z_7	z_8	z_9

- m & n are always odd numbers

Basic Concepts

- Move mask from point to point
- At each point (x, y) , response of filter is calculated using predefined relationship

Example: Linear Spatial Filter

Response, $R = w_1 z_1 + w_2 z_2 + w_3 z_3 + \dots + w_{nm} z_{nm}$

or

$$= \sum_{i=0}^{nm} w_i z_i$$

Basic Concepts

- What happens when centre of filter approaches edge of image ?

Smoothing Spatial Filters

Smoothing Spatial Filters

- Used for smoothing (blurring) and noise deduction

Common Types

1. Smoothing Linear Filters
2. Order-Statistic Filters

Smoothing Linear Filters

- Output (response) is average of pixels in neighbourhood of filter mask
 - Replace value of every pixel with average gray-level value of surrounding pixels
 - Reduces 'sharp' transitions in gray-level

Smoothing Linear Filters

Mean or Box Filter

- All coefficients are equal

$$\frac{1}{9} \times$$



$$R = \frac{1}{mn} (z_1 + z_2 + z_3 + \dots + z_{mn})$$

or

$$= \frac{1}{mn} \sum_{i=0}^{mn} z_i$$

Smoothing Linear Filters

Weighted Average Filter

- Pixels in neighbourhood are multiplied by different coefficients before averaging

$$\frac{1}{16} \times$$



$$R = \frac{(w_1 z_1 + w_2 z_2 + w_3 z_3 + \dots + w_{mn} z_{mn})}{(w_1 + w_2 + w_3 + \dots + z_{mn})}$$

or

$$= \frac{\sum_{i=0}^{mn} w_i z_i}{\sum_{i=0}^{mn} w_i}$$

Smoothing Linear Filters



3 X 3



5 X 5



9 X 9



15 X 15



35 X 35

Order-Statistic Filters

- Non-linear spatial filters
- Response based on order (ranking) of pixel gray-levels within mask
- Centre value (x, y) replaced with this value

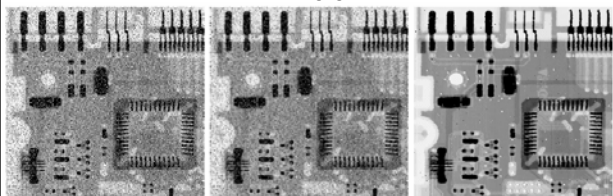
Order-Statistic Filters

- Example: Median Filter
- Replaces value of x, y with median gray-level in mask
- Used for removing certain types of noise

Order-Statistic Filters

3 X 3 averaging filter

3 X 3 median filter



Sharpening Spatial Filters

Sharpening Spatial Filters

- Used to highlight fine detail or enhance detail that has been blurred
- Smoothing - Averaging
- Sharpening - Opposite to averaging ?

Sharpening Spatial Filters

- Enhances regions with high rate of change in gray-scale
 - High intensity gradient (edges)
- De-emphasise areas with slowly varying gray-levels
 - Low intensity gradient (constant intensity)

Derivative Filters

- Averaging is analogous to integration and causes blurring
- Differentiation is expected to have opposite results and sharpen an image.

Derivatives

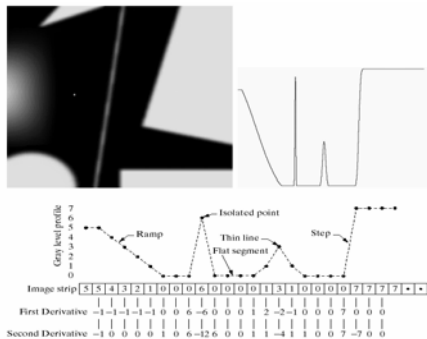
- First derivative

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

- Second derivative

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

Derivatives



Digital Function Derivatives

- First derivative:
 - 0 in constant gray segments
 - Non-zero at the onset of steps or ramps
 - Non-zero along ramps
- Second derivative:
 - 0 in constant gray segments
 - Non-zero at the onset and end of steps or ramps
 - 0 along ramps of constant slope.

Observations

- 1st order derivatives produce thicker edges in an image
- 2nd order derivatives have stronger response to fine detail
- 1st order derivatives have stronger response to a gray level step
- 2nd order derivatives produce a double response at step changes in gray level
- 2nd order derivatives have stronger response to a line than to a step and to a point than to a line

Derivative Filters

- Many different types of sharpening filters
- Example: Gradient Filter

Sharpening Filters

- The filter should have positive coefficients near the center and negative in the outer periphery:

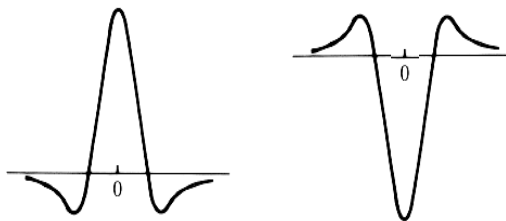
$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

$$\frac{1}{9} \times \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

$$\frac{1}{9} \times \begin{bmatrix} \square & \square & \square \\ \square & -8 & \square \\ \square & \square & \square \end{bmatrix}$$

Sharpening Filters

- Cross section of sharpening filter:



Sharpening Filters

- The sum of the coefficients is 0
 - When the filter is passed over regions of constant gray level, the output of the mask is 0.
- Some scaling and/or clipping is involved (compensates for negative gray-levels)

Gradient Enhancement

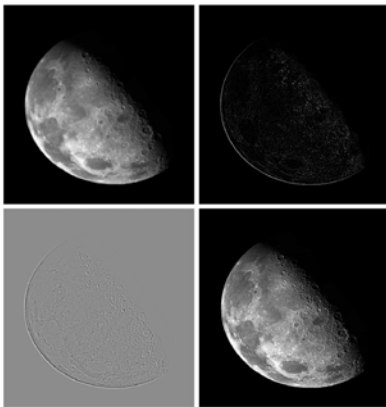
- Highlight gray-level discontinuities
- Deemphasise regions with slowly varying gray
- Produce images with greyish lines on a dark, featureless background
- Background can be recovered while preserving sharpening effect by adding original to derivative image

$$g(x, y) = f(x, y) + \nabla f(x, y)$$

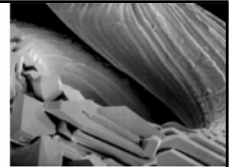
$$g(x, y) = f(x, y) - \nabla f(x, y)$$

Example

$$g(x, y) = f(x, y) + \nabla f(x, y)$$

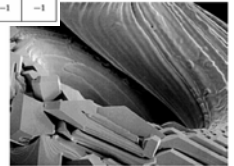
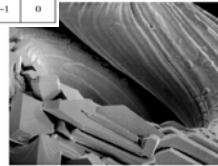


Example



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

-1	-1	-1
-1	9	-1
-1	-1	-1



Combining Spatial Enhancement Methods

- Enhancement usually requires application of several complementary techniques

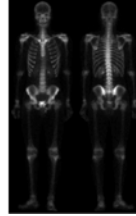
Example

- Nuclear whole body bone scan
 - Detect disease, infection & tumours
- Objective:
 - Bring out more skeletal detail



Example

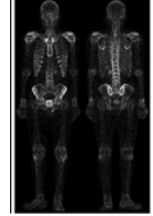
$f(x, y)$



$\nabla f(x, y)$



$g(x, y) = f(x, y) + \nabla f(x, y)$



Summary

- ★Recap: Spatial Enhancement
- ★Spatial Filtering
- ★Smoothing Filters
- ★Sharpening Filters