

Overview (1):

Before We Begin

- Administrative details
- $\ensuremath{\,\,^{\scriptscriptstyle \ensuremath{\mathbb{R}}}}$ Review \rightarrow some questions to consider

Spatial Filtering

- Introduction \rightarrow Review
- Basics of 1D and 2D spatial filtering

Smoothing Spatial Filters

- Introduction
- Smoothing linear filters

Before We Begin

Administrative Details (1):

Lab Four

- Should be a rather straightforward lab to complete
- A lab report is required for this lab and is due in two weeks (e.g., February 28 2006)

Reading Week

- ${\ensuremath{\, \circ \,}}$ Next week ${\ensuremath{\, \rightarrow \,}}$ no lecture or lab
- Test 1
 - \bullet February 28 2006 \rightarrow arrive on time as test will begin at 8:10am sharp!
 - No consideration given if you are late!

Some Questions to Consider (1):

- What is a linear operator and a non-linear operator ?
- How do we show an operator is linear or non-linear ?
- What is the spatial domain ?
- What is image enhancement in the spatial domain ?
- What is spatial filtering?
- What is a smoothing spatial filter ?
- What is an averaging filter ?
- What is a histogram ?
- a Can you think of how to enhance an image by using its histogram ?

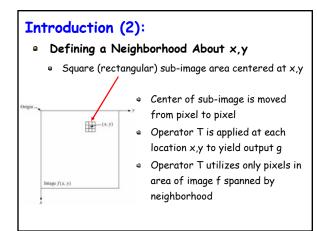
Introduction to Spatial Domain Filtering

Introduction (1):

- Spatial Domain
 - The aggregate of pixels comprising an image
 - Spatial Domain Methods
 - Procedures that operate directly on the image pixels
 - Denoted by the following expression

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

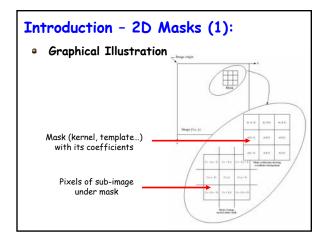
- $f(x,y) \rightarrow input image$
- $g(x,y) \rightarrow output image$
- * $T \rightarrow operator$ defined over some neighborhood of (x,y)

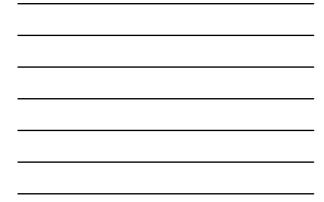


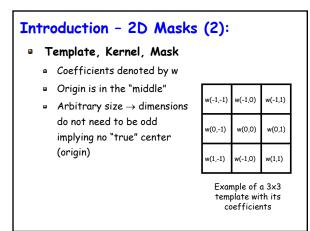
Introduction (3):

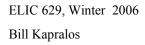
- Defining a Neighborhood About x,y (cont.)
 - Square (rectangular) sub-image area centered at x,y
 - Also known as a mask, template, filter or window
 - Each entry of the sub-image has its own value \rightarrow each value known as a coefficient
 - Mask does not always have to be two-dimensional
 - Can also have one-dimensional mask (more later...)

Basics of Spatial Filtering









Introduction - 2D Masks (3):

- "Mechanics" of Spatial Filtering
 - Moving the template over each pixel of the image
 - At each pixel (x,y) the response (e.g., output value) is determined using some pre-defined relationship
 - For linear spatial filtering, response is given by a sum of the products of the filter coefficients (denoted by w) and the corresponding image pixels "under" the area of the template
 - Mathematically, response g at (x,y) given as g(x,y) = w(-1,-1)f(x-1,y-1) + w(-1,0)f(x-1,y) + ... + w(0,0)f(x,y) + ... + w(1,0)f(x+1,y) + w(1,1)f(x+1,y+1)

Introduction 2D - Masks (4):

- "Mechanics" of Spatial Filtering (cont...)
 - General filtering expression

$$g(x, y) = \sum_{s=-at=-b}^{a} \sum_{w=-b}^{b} w(s,t) f(x+s, y+t)$$

- $m \rightarrow row dimension$
- $n \rightarrow column dimension$
- m = 2a+1 and n = 2b+1 and a,b are non-negative integers or a = (m-1)/2 and b = (n-1)/2
- But this gives output for one pixel location (x,y) only!

Introduction - 2D Masks (5):

"Mechanics" of Spatial Filtering (cont...)

- To generate complete output image, the above equation (process) must be applied to each pixel of input image e.g., for each x = 0 - M-1 and y = 0 - N-1 where M,N are the number of rows and columns of the input image
- Similar to a frequency domain concept called convolution (more on this in the future...)
 - Hence sometimes referred to as "convolving a mask with an image" and the mask is often called a convolution mask



 $R = w_1 z_1 + w_2 z_2 + ... + w_{mn} z_{mn} +$

• "Mechanics" of Spatial Filtering (cont...)

 When we are not interested in processing entire image but rather only a particular pixel (x,y) we can use the following (shorter) mathematical definition:

$$= \sum_{i=1}^{mn} W_i Z_i$$

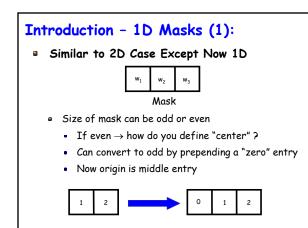


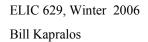
 \rightarrow where, the w's are the filter coefficients and z's are the image gray-levels corresponding to the coefficients and mn is total number of coefficients in template

Introduction - 2D Masks (7):

Important Considerations

- What happens when kernel "placed over" a border pixel ?
- Several approaches for dealing with this
 - Ignore (don't handle) border pixels altogether or any pixels which lead to kernel (or portions of the kernel) "falling" out of image range
 - 2. "Wrap-around"
 - 3. "Zero-pad"
- Keep in mind, some approaches may lead to an output image whose size is not equal to input image!





Smoothing Spatial Filters

Introduction (1):

Purpose of Smoothing Spatial Filters

- Used for blurring, particularly in pre-processing
 - Removal of small detail prior to extraction of large object(s) in image
 - Bridging ("closing") of small gaps in lines or curves
- Also used for noise reduction
 - Can be achieved with a linear or non-linear filter

Smoothing Linear Filters (1):

Essentially an Averaging Filter

- Output of smoothing filter is simply the average of pixels in the neighborhood of filter mask (kernel)
- Also known as a low pass filter
 - Eliminates high frequency components (we will describe this further in later lectures)
- Idea of smoothing filter
 - Random noise typically consists of sharp transitions in gray levels

Smoothing Linear Filters (2):

- Idea of smoothing filter (cont...)
 - By replacing the value of every pixel by the average of its neighbors, we essentially reduce the sharp transitions in gray levels
- Most obvious application of a smoothing filter is noise reduction
- However, beware!
 - Not all sharp transitions are bad and un-wanted!
 - Edges which are typically very important and wanted features of an image are also defined as sharp transitions in gray levels

Smoothing Linear Filters (3):

- However, beware! (cont...)
 - Since smoothing filter removes sharp transitions in gray level, averaging (smoothing) filters blur edges!
- Several other applications of smoothing (averaging) filters in addition to noise reduction
 - Smoothing of false contours which result when not using a large number of gray levels
 - Removing irrelevant details in an image → pixel regions that are small in comparison to the size of the filter kernel

Smoothing Linear Filters (4):

Smoothing Filter Kernels

Example of a 3 x 3 smoothing filter

$$\frac{1}{9} \times \frac{1}{1} \frac{1}{1} \frac{1}{1} \frac{1}{1} \frac{1}{1} R = \frac{1}{9} \sum_{i=1}^{9} Z^{i}$$

- Above kernel (filter) produces average of the pixels under the mask
- Notice that coefficients are equal to 1 and not 1/9!
 - One division instead of nine \rightarrow more efficient!



- This is an example of a weighted average filter
 - Coefficients are not all the same value!
 - Some pixels multiplied by higher values thus giving those pixels more importance in average

Smoothing Linear Filters (6):

• Another example of a 3 x 3 smoothing filter (cont...)



- Center coefficient is highest meaning center pixel is given most importance
- Other coefficients are reduced inversely as a function of distance from the center coefficient
 - Diagonal terms are further away than the "edge" neighbors and thus the corresponding pixels in image provide less importance to average

Smoothing Linear Filters (7):

- Another example of a 3 x 3 smoothing filter (cont...)
 - Many other types of coefficient masks are also available depending on application but typically try to keep the sum of the coefficients an integral power of 2 (e.g., 16 as in previous example)
- In general, hard to notice differences between images filtered by both these filter examples

Smoothing Linear Filters (8):

General Filter Implementation

 Recall the filtering expression for filtering M x N image with weighted averaging filter of size m x n

$$g(x, y) = \sum_{s=-at=-b}^{a} w(s,t) f(x+s, y+t)$$

 The general expression equation can now be stated as (again, given an N x M image with m x n filter where m,n are odd)

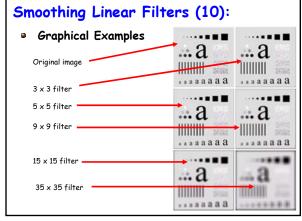
$$g(x, y) = \frac{\sum_{s=-at=-b}^{a} \sum_{s=-at=-b}^{b} w(s, t) f(x + s, y + t)}{\sum_{s=-as=-b}^{a} \sum_{s=-b}^{b} w(s, t)}$$

Smoothing Linear Filters (9):

• General Filter Implementation (cont...)

$$g(x, y) = \frac{\sum_{s=-at=-b}^{a} w(s, t) f(x + s, y + t)}{\sum_{s=-as=-b}^{a} w(s, t)}$$

- Complete filtered image is obtained by applying above equation for each x = 0,1,2, ..., M-1 and y = 0,1,2, ..., N-1
- Denominator is sum of the mask (kernel) coefficients and is constant (e.g., computed once!)
 - Typically division applied once to output image rather than at each stage (pixel output)





Smoothing Linear Filters (11):

- Some Notes Regarding Averaging Filters
 - For small filter (e.g., 3 x 3) a slight general blurring of entire image occurs but details that are about same size of filter are affected considerably
 - Noise is less pronounced
 - Jagged borders are "pleasantly" smoothed
 - As filter size increases, blurring is more pronounced

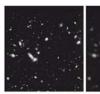
Smoothing Linear Filters (12):

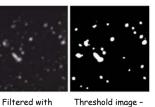
- **Image Blurring** ٠
 - Important application of averaging filter is blurring to get "gross representation" of objects of interest
 - Smaller objects blend into the background
 - Larger objects become more "blob-like" and easy to detect
 - Size of mask (kernel) determines size of objects to be blended into background \rightarrow smaller mask, smaller objects blended to background

Smoothing Linear Filters (13):

- Image Blurring Graphical Example
 - Image obtained with Hubble space telescope

mask





Original image

Threshold image -15 x 15 averaging small objects have disappeared

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Order-Statistics Filters (1):

Non-linear Spatial Filters

- Response of filter is based on ordering (ranking) the pixels contained in image area encompassed by filter and then replacing center pixel with value determined by ranking result
- One of the most known examples is the median filter

Order-Statistics Filters (2):

Median Filter

- Replaces the value of a pixel with median of the gray levels in the neighborhood of that pixel
- Can provide excellent noise reduction with less blurring than averaging filters
 - Particularly effective for impulsive noise also known as salt-and-pepper noise due to its appearance as white and black dots in the image
- What is the median ?
 - Given a set of values, the median ξ of the set is chosen such that half the values of set are less than ξ and half are more

Order-Statistics Filters (3):

Median Filter (cont...)

- To perform median filtering:
 - 1. Sort the values of the pixel in question at spatial location (x,y) and its neighbors
 - 2. Determine the median value $\boldsymbol{\xi}$
 - 3. Assign intensity value at location (x,y) the median value
- Given 3 x 3 mask \rightarrow median value is the 5th largest value, 5 x 5 mask \rightarrow median is 13th largest value

Order-Statistics Filters (4):

Median Filter (cont...)

- Principle function of median filter is to force points with distinct gray levels to be more like their neighbors
 - Isolated clusters of pixels that are light or dark with respect to their neighbors are eliminated

Order-Statistics Filters (5):

• Median Filter Graphical Example

Comparison between averaging filter







Original image corrupted by saltand-pepper noise Image filtered withImage filtered with3 x 3 average filter3 x 3 median filter

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