

Image Filtering

COSC342

Lecture 3
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In this Lecture

- ▶ Image filters
- ▶ Kernels and convolution
- ▶ Blurring – mean filter, Gaussian filter
- ▶ Edge detection – Sobel filters
- ▶ Separable filters

Image Filters

- ▶ A filter takes an image and applies some operation at every pixel
 - ▶ The same operation is applied at all pixels
 - ▶ This operation produces a new value for each pixel
 - ▶ Often this is viewed as an image itself, which is the result of the filter
- ▶ We will consider just greyscale images
 - ▶ Can apply filters independently to RGB colour channels
 - ▶ In HSV etc, might just apply filters to the intensity channel
- ▶ Many filters are applied in a process called *convolution*

Image Convolution

- ▶ We have a kernel, K , which is a $2k + 1 \times 2k + 1$ matrix of numbers
- ▶ This is applied at each point, (x, y) , in an image, I as follows:

$$(I * K)(x, y) = \sum_{i=-k}^k \sum_{j=-k}^k K(i, j) I(x - i, y - j)$$

- ▶ This is called discrete *convolution*
 - ▶ The kernel moves over the image, and is applied to its centre pixel
 - ▶ We multiply corresponding kernel and image entries
 - ▶ The result is the sum of these products

Image Convolution

1	1	1	2	4	5	6	7
1	3	2	4	5	7	7	8
2	1	3	5	6	6	8	8
3	2	1	3	2	6	7	8
4	3	2	3	1	4	6	7
5	4	4	5	4	2	3	6
5	6	7	8	7	6	5	6
7	8	8	8	8	8	7	5

0	1/8	0
1/8	1/2	1/8
0	1/8	0

$$\begin{aligned}(I * K)(5, 4) &= (0 \times 5) + \left(\frac{1}{8} \times 6\right) + (0 \times 6) + \left(\frac{1}{8} \times 3\right) + \left(\frac{1}{2} \times 2\right) + \dots \\ &= \frac{0 + 6 + 0 + 3 + 8 + 6 + 0 + 3 + 0}{8} = \frac{24}{8} = 3\end{aligned}$$

Convolution Problems

- ▶ Some filters give negative outputs, fractions, or go out of range
 - ▶ Can round values, or change type of image data, etc.
 - ▶ Can use mid-grey as zero, white as positive, black as negative
- ▶ All filters have problems at the edges of images

?	?	?	?	?	?	?		
?	1	1	1	2	4	5	6	7
?	1	3	2	4	5	7	7	8
	2	1	3	5	6	6	8	7

- ▶ Can ignore the edges (shrinking the output image a bit)
- ▶ Can predict the values (nearest neighbour, linear models, etc.)
- ▶ Can tile or mirror the image

Blurring Filters: Mean Filters

- ▶ Mean filters take a simple average over a window
- ▶ There is a family of them depending on size – 3×3 , 5×5 , etc.

$$M_{3 \times 3} = \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \quad M_{5 \times 5} = \frac{1}{25} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

- ▶ These blur and smooth the image

Mean Filters



$$*M_{3 \times 3} =$$



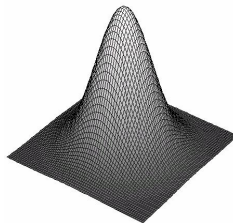
$$*M_{7 \times 7} =$$



More Blurring: Gaussian Filters

- ▶ Mean filters treat all pixels equally
- ▶ There are good reasons to give more weight to pixels nearer the centre
- ▶ A common way to do this is with Gaussian filters

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



- ▶ The parameter σ determines how much blurring we get
 - ▶ Beyond about 3σ the values are very close to zero
 - ▶ So we can use a roughly $6\sigma \times 6\sigma$ kernel
 - ▶ Need to normalise kernel values so that they add to 1

Gaussian Filters



$$*G_{\sigma=1} =$$



$$*G_{\sigma=3} =$$



Edge Detection: Sobel Filters

- ▶ Another common use of filters is in *edge detection*
- ▶ An edge in an image is a place where the intensity changes quickly
- ▶ The Sobel filters detect edges in the x - and y -directions:

$$S_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad S_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

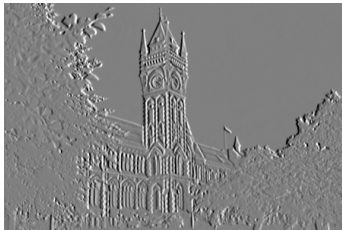
- ▶ These can be combined to make a gradient vector, $\mathbf{g} = \begin{bmatrix} S_x \\ S_y \end{bmatrix}$

Sobel Filters

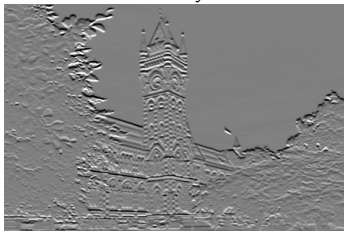
I



$I * S_x$



$I * S_y$



$|g|$



Separable Filters

- ▶ If we have an image with n pixels, applying a $k \times k$ filter is $O(nk^2)$
- ▶ This can get expensive with large filters
 - ▶ $G_{\sigma=3}$, for example is 19×19 – 361 multiplies and additions per pixel
- ▶ Many 2D filters can be separated into two 1D filters
 - ▶ First we apply a $1 \times k$ filter, F_x , to each pixel
 - ▶ Next we apply a $k \times 1$ filter, F_y , to each pixel
 - ▶ This has the same effect as applying the 2D filter ($F_y F_x$)

Separable Filters

$$M_{3 \times 3} = \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = \left(\frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right) \left(\frac{1}{3} [1 \quad 1 \quad 1] \right)$$

$$G_{\sigma=\frac{1}{2}} \approx \begin{bmatrix} 0.04 & 0.12 & 0.04 \\ 0.12 & 0.36 & 0.12 \\ 0.04 & 0.12 & 0.04 \end{bmatrix} = \begin{bmatrix} 0.2 \\ 0.6 \\ 0.2 \end{bmatrix} [0.2 \quad 0.6 \quad 0.2]$$

$$S_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} [-1 \quad 0 \quad 1]$$

Other Filters

- ▶ Sharpening filter



$$* \begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix} =$$



- ▶ Emboss filter



$$* \begin{bmatrix} -2 & -1 & 0 \\ -1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} =$$



- ▶ Are these separable?

The Median Filter

- ▶ Not all filters can be implemented as convolution
- ▶ A common filter is the Median filter:
 - ▶ Sort all of the pixel values in a $k \times k$ window
 - ▶ Choose the middle value as the result



→
5×5



Tutorials and Labs

- ▶ This week's tutorial:
 - ▶ Matrix and vector mathematics
 - ▶ Have a go *before the tutorial*
 - ▶ Come along if you have any questions or issues
- ▶ Monday's Lab:
 - ▶ Matrices in C++ / OpenCV introduction