

# CS540 Introduction to Artificial Intelligence

## Lecture 7

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Based on lecture slides by Jerry Zhu and Yingyu Liang

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# Description of Algorithm

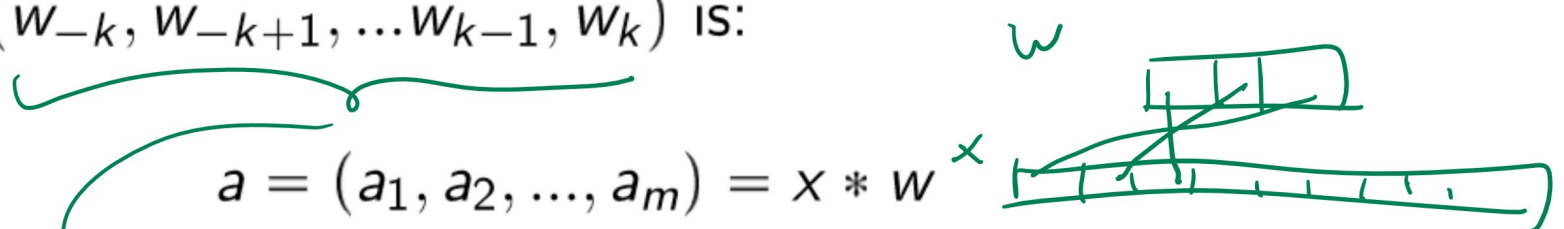
## Description

- Convolve the input image with a filter.
- Pool the output of convolution.
- Feed the output of pooling into a neural network.

# One Dimensional Convolution

## Definition

- The convolution of a vector  $x = (x_1, x_2, \dots, x_m)$  with a filter  $w = (w_{-k}, w_{-k+1}, \dots, w_{k-1}, w_k)$  is:



$$a = (a_1, a_2, \dots, a_m) = x * w$$

$$a_j = \sum_{t=-k}^k w_t x_{j-t}, j = 1, 2, \dots, m$$

linear combination  $a_j = \sum_{t=-k}^k w_t x_t$

- $w$  is also called a kernel (different from the kernel for SVMs).
- The elements that do not exist are assumed to be 0.









# Convolution Example, Part II

## Quiz (Graded)

Q4

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

*(Handwritten annotations: a green box around the first two columns of the first matrix, a blue box around the first two rows of the second matrix, and a red 'X' below the first matrix.)*

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

*(Handwritten annotations: a green box around the matrix, a red 'w' below it, and a red 'w'' below the second row.)*

- A:  $\begin{bmatrix} -1 & -3 & -3 \\ 0 & 0 & 0 \\ 1 & 3 & 3 \end{bmatrix}$ , B:  $\begin{bmatrix} -3 & -3 & 3 \\ -4 & -4 & 4 \\ -3 & -3 & 3 \end{bmatrix}$

*(Handwritten annotations: a green circle around the '0' in the bottom-right of matrix A.)*

- C:  $\begin{bmatrix} -3 & -4 & -3 \\ -3 & -4 & -3 \\ 3 & 4 & 3 \end{bmatrix}$ , D:  $\begin{bmatrix} -1 & 0 & 1 \\ -3 & 0 & 3 \\ -3 & 0 & 3 \end{bmatrix}$

- E: none of the above

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$$\begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

*(Handwritten annotations: a blue box around the vector, a green circle around the 0, and a blue circle around the 3.)*

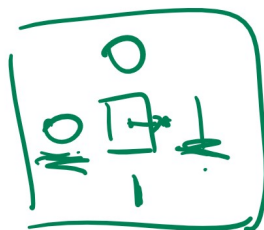
# Image Gradient

## Definition

- The gradient of an image is defined as the change in pixel intensity due to the change in the location of the pixel.

$$\frac{\partial I(x, y)}{\partial x} \approx \frac{I\left(x + \frac{\varepsilon}{2}, y\right) - I\left(x - \frac{\varepsilon}{2}, y\right)}{\varepsilon}, \varepsilon = 1$$

$$\frac{\partial I(x, y)}{\partial y} \approx \frac{I\left(x, y + \frac{\varepsilon}{2}\right) - I\left(x, y - \frac{\varepsilon}{2}\right)}{\varepsilon}, \varepsilon = 1$$

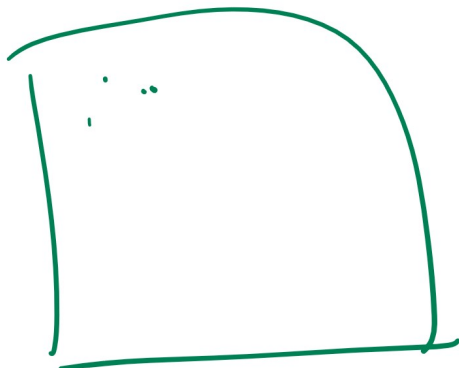


# Image Derivative Filters

## Definition

- The gradient can be computed using convolution with the following filters.


$$w_x = \begin{bmatrix} 0 & 0 & 0 \\ -1 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}, w_y = \begin{bmatrix} 0 & -1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$



# Sobel Filter

## Definition

- The Sobel filters also are used to approximate the gradient of an image.


$$W_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}, W_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

# Decomposition of Filters

## Definition

- The Sobel filters can be decomposed into two one dimensional filters.

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

- It is significantly faster to do two one dimensional convolutions than to do one two dimensional convolution.

# Gradient of Images

## Definition

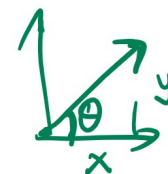
- The gradient of an image  $I$  is  $(\nabla_x I, \nabla_y I)$ .

$$\nabla_x I = W_x * I, \nabla_y I = W_y * I$$

- The gradient magnitude is  $G$  and gradient direction  $\Theta$  are the following.

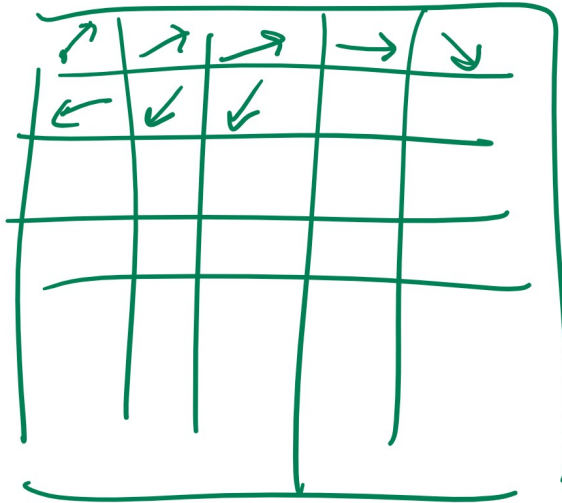
$$\underline{G} = \sqrt{\nabla_x^2 + \nabla_y^2} \quad \text{done for each pixel!}$$

$$\Theta = \arctan \left( \frac{\nabla_y}{\underline{\nabla_x}} \right)$$



# Gradient of Images Diagram

Definition

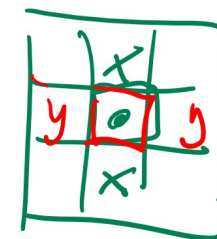




# Laplacian of Image

## Definition

- The Laplacian of an image  $I$  is defined as the sum of the second derivatives.



$$\nabla^2 I(x, y) = \frac{\partial^2 I(x, y)}{\partial x^2} + \frac{\partial^2 I(x, y)}{\partial y^2}$$

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

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

# Laplacian Filter

## Definition

- The Laplacian can be computed using convolution with the following filters.

$$W_L = \begin{bmatrix} 0 & 0 & 0 \\ 1 & -2 & 1 \\ 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 1 & 0 \\ 0 & -2 & 0 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\nabla^2 I = W_L * I$$

# Edge Detection

## Discussion

- Both the gradient and Laplacian of an image can be used to find edge pixels in an image.
- Images usually contain noise. The noises are not edges and are usually removed before computing the gradient.



# 1 Dimensional Gaussian Filter

## Definition

- The Gaussian filter can be decomposed into two one dimensional filters as well.

$$W_\sigma = w_\sigma * w_\sigma, (w_\sigma)_t = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{t^2}{2\sigma^2}\right)$$

# Gaussian Filter Example 3

## Definition

- When filter size  ~~$k=$~~   <sup>$k=1$</sup>  3, and standard deviation  $\sigma = 0.8$ :

$$W_{\sigma} = \frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

- Sobel filter is approximately the combination of the gradient filter and the Gaussian filter.

# Laplacian of Gaussian

## Definition

- The Laplacian filter and the Gaussian filter are usually also combined into one filter called Laplacian of Gaussian filter (LoG filter).

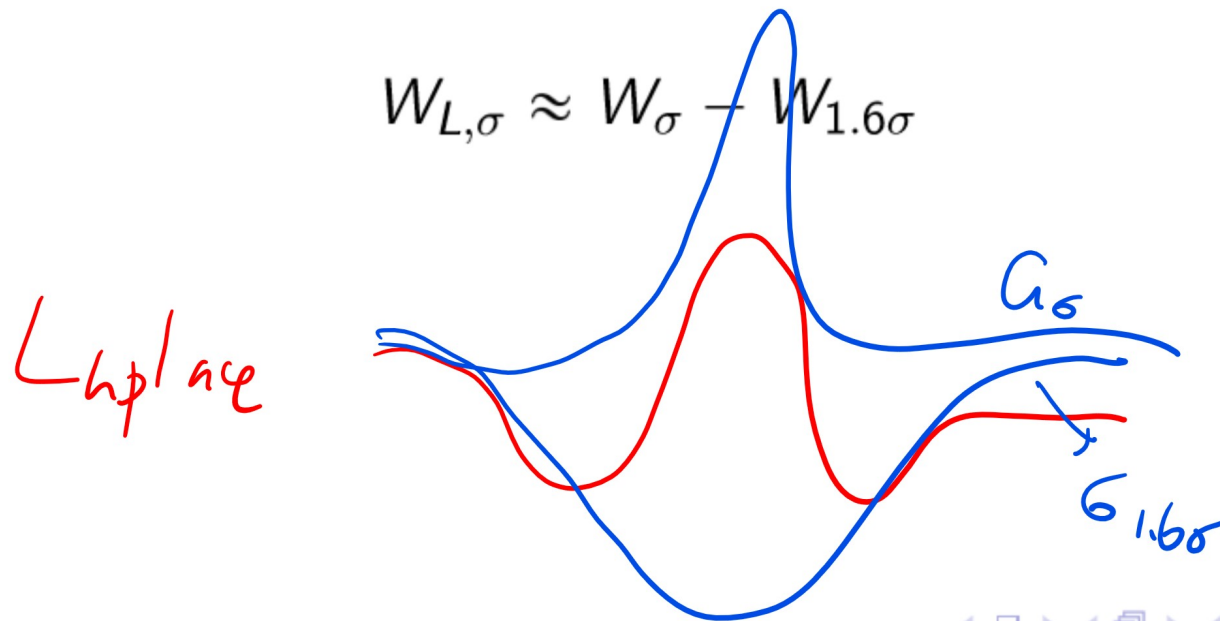
$$W_{L,\sigma} : (W_{L,\sigma})_{t,t'} = -\frac{1}{\pi\sigma^4} \left( 1 - \frac{t^2 + t'^2}{2\sigma^2} \right) \exp \left( -\frac{t^2 + t'^2}{2\sigma^2} \right)$$

→ cannot decompose

# Difference of Gaussian

## Definition

- The Laplacian of Gaussian filter is difficult to compute because it cannot be decomposed into two one dimensional filters. Therefore an approximation is used called the Difference of Gaussian filter (DoG filter).







# Image Pyramids

## Discussion

- There are edges at different scales of the image. Images are blurred and downsampled to get images with different scales.
- An image pyramid contains images at scales  $1, \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots$

# Feedback, Part I

Admin

- Overall, the pace of the course is:
- Not anonymous. Not required to answer.
- A: Too fast
- B: Too slow
- C: Just right

# Feedback, Part II

Admin

- Overall, the assigned homeworks (math + quiz) are:
- Not anonymous. Not required to answer.
- A: Too hard
- B: Too easy
- C: Just right

# Feedback, Part III

Admin

- Overall, the assigned homeworks (programming) are:
- Not anonymous. Not required to answer.
- A: Too hard
- B: Too easy
- C: Just right

# Feedback, Anonymous

Admin

- You can login Room CS540S0 to give additional anonymous feedback.
- No more quiz questions in the second half of the lecture.

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