# CS540 Introduction to Artificial Intelligence Lecture 11

Young Wu

Based on lecture slides by Jerry Zhu, Yingyu Liang, and Charles Dyer

June 27, 2022

# Coordination Game Quiz



## Exam Date







# Review Session



### One Dimensional Convolution Definition

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

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

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

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

### Two Dimensional Convolution

• The convolution of an  $m \times m$  matrix X with a  $(2k+1) \times (2k+1)$  filter W is:

$$A = X * W$$

$$A_{j,j'} = \sum_{s=-k}^{k} \sum_{t=-k}^{k} W_{s,t} X_{j-s,j'-t}, j, j' = 1, 2, ..., m$$

- The matrix W is indexed by (s,t) for s=-k,-k+1,...,k-1,k and t=-k,-k+1,...,k-1,k.
- The elements that do not exist are assumed to be 0.

### Convolution Diagram and Demo

### 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.

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

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

### Image Derivative Filters Definition

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

$$w_{x} = \begin{bmatrix} -1 & 0 & 1 \end{bmatrix}, w_{y} = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

#### Sobel Filter

Definition

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

$$W_{x} = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}, W_{y} = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

#### 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.

$$G = \sqrt{\nabla_x^2 + \nabla_y^2}$$

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

## Gradient of Images Demo

### Convolution Example

# Gradient Example Quiz

# Convolution Example 1

#### SIFT Discussion

 Scale Invariant Feature Transform (SIFT) features are features that are invariant to changes in the location, scale, orientation, and lighting of the pixels.

#### Classification

#### Discussion

- SIFT features are not often used in training classifiers and more often used to match the objects in multiple images.
- HOG features are usually computed for every cell in the image and used as features (in place of pixel intensities) in classification algorithms such as SVM.