# CS540 Introduction to Artificial Intelligence Lecture 3

Young Wu

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

July 1, 2022

# Two-thirds of the Average Game

- Pick an integer between 0 and 100 (including 0 and 100) that is the closest to two-thirds of the average of the numbers other people picked.
- The results from the previous lecture is posted on the *Q*1 page of the course website.

# AND Operator Data

Sample data for AND

<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	у
0	0	0
0	1	0
1	0	0
1	1	1

### Learning AND Operator

• Which one of the following is AND?

• 
$$A: \hat{y} = \mathbb{1}_{\{1x_1+1x_2-1.5 \ge 0\}}$$

• 
$$B: \hat{y} = \mathbb{1}_{\{1x_1+1x_2-0.5\geqslant 0\}}$$

• 
$$C: \hat{y} = \mathbb{1}_{\{-1x_1+0.5 \ge 0\}}$$

• 
$$D: \hat{y} = \mathbb{1}_{\{-1x_1-1x_2+0.5 \geq 0\}}$$

• E : None of the above

### OR Operator Data Quiz

Sample data for OR

<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	У
0	0	0
0	1	1
1	0	1
1	1	1

## Learning OR Operator

• Which one of the following is OR?

• 
$$A: \hat{y} = \mathbb{1}_{\{1x_1+1x_2-1.5 \ge 0\}}$$

• 
$$B: \hat{y} = \mathbb{1}_{\{1x_1 + 1x_2 - 0.5 \ge 0\}}$$

• 
$$C: \hat{y} = \mathbb{1}_{\{-1x_1+0.5 \ge 0\}}$$

• 
$$D: \hat{y} = \mathbb{1}_{\{-1x_1 - 1x_2 + 0.5 \ge 0\}}$$

• E : None of the above

## Learning XOR Operator

• Which one of the following is XOR?

• 
$$A: \hat{y} = \mathbb{1}_{\{1x_1+1x_2-1.5 \ge 0\}}$$

• 
$$B: \hat{y} = \mathbb{1}_{\{1x_1 + 1x_2 - 0.5 \ge 0\}}$$

• 
$$C: \hat{y} = \mathbb{1}_{\{-1x_1+0.5 \ge 0\}}$$

• 
$$D: \hat{y} = \mathbb{1}_{\{-1x_1 - 1x_2 + 0.5 \ge 0\}}$$

• E : None of the above

## Learning XOR Operator Network

•  $y = x_1 \text{ XOR } x_2 \text{ is the same as}$  $y = (x_1 \text{ OR } x_2) \text{ AND } (x_1 \text{ NAND } x_2)$ 

### Single Layer Perceptron Motivation

- Perceptrons can only learn linear decision boundaries.
- Many problems have non-linear boundaries.
- One solution is to connect perceptrons to form a network.

### Decision Boundary Diagram Motivation

### Multi-Layer Perceptron

Motivation

• The output of a perceptron can be the input of another.

$$a = g\left(w^{T}x + b\right)$$

$$a' = g\left(w'^{T}a + b'\right)$$

$$a'' = g\left(w''^{T}a' + b''\right)$$

$$\hat{y} = \mathbb{1}_{\{a'' > 0\}}$$

### Neural Network Biology Motivation

- Human brain: 100,000,000,000 neurons.
- Each neuron receives input from 1,000 others.
- An impulse can either increase or decrease the possibility of nerve pulse firing.
- If sufficiently strong, a nerve pulse is generated.
- The pulse forms the input to other neurons.

### Theory of Neural Network

- In theory:
- 1 Hidden-layer with enough hidden units can represent any continuous function of the inputs with arbitrary accuracy.
- 2 Hidden-layer can represent discontinuous functions.
  - In practice:
- AlexNet: 8 layers.
- @ GoogLeNet: 27 layers (or 22 + pooling).
- ResNet: 152 layers.

### Gradient Descent

- The derivatives are more difficult to compute.
- The problem is no longer convex. A local minimum is no longer guaranteed to be a global minimum.
- Need to use chain rule between layers called backpropagation.

#### • Initialize random weights.

- (Feedforward Step) Evaluate the activation functions.
- (Backpropagation Step) Compute the gradient of the cost function with respect to each weight and bias using the chain rule.
- Update the weights and biases using gradient descent.
- Repeat until convergent.

### Neural Network Demo

# Two-Layer Neural Network Weights Diagram 1 Motivation

# Two-Layer Neural Network Weights Diagram 2 Motivation

# Two-Layer Neural Network Weights Diagram 3 Motivation

### Gradient Step, Combined

#### Definition

 Put everything back into the chain rule formula. (Please check for typos!)

$$\frac{\partial C}{\partial w_{j'j}^{(1)}} = \sum_{i=1}^{n} (a_i - y_i) a_i (1 - a_i) w_j^{(2)} a_{ij}^{(1)} \left( 1 - a_{ij}^{(1)} \right) x_{ij'}$$

$$\frac{\partial C}{\partial b_j^{(1)}} = \sum_{i=1}^{n} (a_i - y_i) a_i (1 - a_i) w_j^{(2)} a_{ij}^{(1)} \left( 1 - a_{ij}^{(1)} \right)$$

$$\frac{\partial C}{\partial w_j^{(2)}} = \sum_{i=1}^{n} (a_i - y_i) a_i (1 - a_i) a_{ij}^{(1)}$$

$$\frac{\partial C}{\partial b^{(2)}} = \sum_{i=1}^{n} (a_i - y_i) a_i (1 - a_i)$$

### Gradient Descent Step

#### Definition

 The gradient descent step is the same as the one for logistic regression.

$$\begin{split} & w_{j}^{(2)} \leftarrow w_{j}^{(2)} - \alpha \frac{\partial C}{\partial w_{j}^{(2)}}, j = 1, 2, ...., m^{(1)} \\ & b^{(2)} \leftarrow b^{(2)} - \alpha \frac{\partial C}{\partial b^{(2)}}, \\ & w_{j'j}^{(1)} \leftarrow w_{j'j}^{(1)} - \alpha \frac{\partial C}{\partial w_{j'j}^{(1)}}, j' = 1, 2, ...., m, j = 1, 2, ...., m^{(1)} \\ & b_{j}^{(1)} \leftarrow b_{j}^{(1)} - \alpha \frac{\partial C}{\partial b_{i}^{(1)}}, j = 1, 2, ...., m^{(1)} \end{split}$$

### Learning Logical Operators, XOR

• What function does the multi-layer LTU perceptron network with  $w_{11}^{(1)}=-1, w_{21}^{(1)}=-1, b_1^{(1)}=1.5, w_{12}^{(1)}=1, w_{22}^{(1)}=1, b_2^{(1)}=-0.5, w_1^{(2)}=1, w_2^{(2)}=1, b^{(2)}=-1.5$  compute?

<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	УА	УВ	УС	УD	УЕ
0	0	0	0	1	1	0
0	1	0	1	1	0	1
1	0	0	1	1	0	1
1	1	1	1	0	1	0

# Learning Logical Operators, XOR, Diagram Quiz

<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	УА	УВ	УС	УD	УЕ
0	0	0	0	1	1	0
0	1	0	1	1	0	1
1	0	0	1	1	0	1
1	1	1	1	0	1	0

# Learning Logical Operators, XOR, Answer

## Three-Layer Neural Network Weights Diagram Motivation