# CS540 Introduction to Artificial Intelligence Lecture 2

Young Wu

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

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# Two-thirds of the Average Game

# Quizzes, Math Homework, Discussions

## Office Hours, Discussion Sessions

## Supervised Learning Motivation

## Loss Function Diagram Motivation

#### Zero-One Loss Function

#### Motivation

• An objective function is needed to select the "best"  $\hat{f}$ . An example is the zero-one loss.

$$\hat{f} = \underset{f}{\operatorname{argmin}} \sum_{i=1}^{n} \mathbb{1}_{\{f(x_i) \neq y_i\}}$$

- $\operatorname{argmin}_f$  objective (f) outputs the function that minimizes the objective.
- The objective function is called the cost function (or the loss function), and the objective is to minimize the cost.

#### Squared Loss Function

#### Motivation

- Zero-one loss counts the number of mistakes made by the classifier. The best classifier is the one that makes the fewest mistakes.
- Another example is the squared distance between the predicted and the actual y value:

$$\hat{f} = \underset{f}{\operatorname{argmin}} \frac{1}{2} \sum_{i=1}^{n} (f(x_i) - y_i)^2$$

# Loss Functions Equivalence Quiz

# Loss Functions Equivalence, Answer

## Function Space Diagram Motivation

## Hypothesis Space

- There are too many functions to choose from.
- There should be a smaller set of functions to choose  $\hat{f}$  from.

$$\hat{f} = \underset{f \in \mathcal{H}}{\operatorname{argmin}} \frac{1}{2} \sum_{i=1}^{n} (f(x_i) - y_i)^2$$

ullet The set  ${\cal H}$  is called the hypothesis space.

#### **Activation Function**

#### Motivation

• Suppose  $\mathcal H$  is the set of functions that are compositions between another function g and linear functions.

$$\left(\hat{w}, \hat{b}\right) = \underset{w,b}{\operatorname{argmin}} \frac{1}{2} \sum_{i=1}^{n} \left(a_i - y_i\right)^2$$
  
where  $a_i = g\left(w^T x + b\right)$ 

g is called the activation function.

## Linear Threshold Unit

 One simple choice is to use the step function as the activation function:

$$g\left(\boxed{\cdot}\right) = \mathbb{1}_{\left\{\begin{array}{c} \cdot \\ \cdot \\ \end{array} \geqslant 0\right\}} = \left\{\begin{array}{cc} 1 & \text{if } \boxed{\cdot} \geqslant 0 \\ 0 & \text{if } \boxed{\cdot} < 0 \end{array}\right.$$

• This activation function is called linear threshold unit (LTU).

#### Sigmoid Activation Function

Motivation

 When the activation function g is the sigmoid function, the problem is called logistic regression.

$$g\left(\boxed{\cdot}\right) = \frac{1}{1 + \exp\left(-\boxed{\cdot}\right)}$$

• This g is also called the logistic function.

## Sigmoid Function Diagram Motivation

### Cross-Entropy Loss Function

 The cost function used for logistic regression is usually the log cost function.

$$C(f) = -\sum_{i=1}^{n} (y_i \log (f(x_i)) + (1 - y_i) \log (1 - f(x_i)))$$

• It is also called the cross-entropy loss function.

### Logistic Regression Objective

 The logistic regression problem can be summarized as the following.

$$\left(\hat{w}, \hat{b}\right) = \underset{w,b}{\operatorname{argmin}} - \sum_{i=1}^{n} \left(y_{i} \log \left(a_{i}\right) + \left(1 - y_{i}\right) \log \left(1 - a_{i}\right)\right)$$
where  $a_{i} = \frac{1}{1 + \exp\left(-z_{i}\right)}$  and  $z_{i} = w^{T} x_{i} + b$ 

#### Optimization Diagram

Motivation

# Logistic Regression Description

- Initialize random weights.
- Evaluate the activation function.
- Compute the gradient of the cost function with respect to each weight and bias.
- Update the weights and biases using gradient descent.
- Repeat until convergent.

#### Gradient Descent Step

Definition

• For logistic regression, use chain rule twice.

$$w = w - \alpha \sum_{i=1}^{n} (a_i - y_i) x_i$$

$$b = b - \alpha \sum_{i=1}^{n} (a_i - y_i)$$

$$a_i = g\left(w^T x_i + b\right), g\left(\boxed{\cdot}\right) = \frac{1}{1 + \exp\left(-\boxed{\cdot}\right)}$$

 $oldsymbol{lpha}$  is the learning rate. It is the step size for each step of gradient descent.

## Perceptron Algorithm Definition

• Update weights using the following rule.

$$w = w - \alpha (a_i - y_i) x_i$$
  

$$b = b - \alpha (a_i - y_i)$$
  

$$a_i = \mathbb{1}_{\{w^T x_i + b \geqslant 0\}}$$

## Learning Rate Diagram Definition

### Other Non-linear Activation Function

- Activation function:  $g(\boxed{\cdot}) = \tanh(\boxed{\cdot}) = \frac{e^{\boxed{\cdot}} e^{-\boxed{\cdot}}}{e^{\boxed{\cdot}} + e^{-\boxed{\cdot}}}$
- Activation function:  $g( \overline{ } ) = \arctan( \overline{ } )$
- Activation function (rectified linear unit):  $g\left(\boxed{\cdot}\right) = \boxed{1}_{\left\{\boxed{\cdot}\right\} > 0}$
- All these functions lead to objective functions that are convex and differentiable (almost everywhere). Gradient descent can be used.

# Gradient Descent

# Gradient Descent, Answer

# Gradient Descent, Answer Too

# Gradient Descent

# Gradient Descent, Another One, Answer

# Gradient Descent, Another One Too Quiz

# Gradient Descent, Another One Too, Answer

#### Convexity Diagram

Discussion