#### BMEG 3105

Fall 2022

## Data analytics for personalized genomics and precision medicine

## **Course introduction**

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## **Outline of lecture**

#### I. Recap from last lecture

- A. Classification
- B. K-nearest neighbour classification
- C. Clustering VS Classification
- D. Logistic regression

#### II. Today's agenda

- A. Logistic Regression Model Training
  - a) How to train?
    - 1. Standard process
    - 2. Loss Function
    - 3. How to minimize
    - 4. To get the formula
- B. From logistic Regression to Neural Networks
  - a) The simplest neural networks
  - b) From logistic regression to neural network
  - c) From network to deep learning

#### III. Self-study material

## I. Recap from last lecture

### A. Classification

- a) What is classification?
  Given that there is a collection of records (Training set)
  Find a method to assign the class of previously unseen records based on their other attributes and the training set as accurately as possible.
- b) How to do classification
   First, get the training data with class
   Second, build a classification method
   Third, insert the data to be classified
   Then, get the result

#### B. K nearest neighbour classification

Standard process of KNN:

- 1. Normalization
- 2. Computer distances
- 3. Identify the K most similar data
- 4. Take their class out and find the mode class

#### **C.** Clustering VS Classification



#### **D.** Logistic regression



# II. Today's agenda

### A. Logistic Regression Model Training

a) How to train?

- 1. Standard Process
  - Get W<sub>h</sub>, W<sub>w</sub>, and W<sub>0</sub>, (Initialization)
  - Make the model fit the training data
  - Make  $\frac{1}{1+e^{-(w_hH+w_wW+w_0)}} \ge 0.5$  correct for the training data
  - $Y^{output} = \frac{1}{1 + e^{-(w_h H + w_w W + w_0)}}$  (1 for male and 0 for female).
- 2. Loss Function
  - As we want Y<sup>output</sup> should be as close as true data possible.
  - $(Y^{output} Y)^2$  should be as small as possible. (Y is the true data we have for training data)
- 3. How to minimize
  - Use differentiation to find the minimum, when  $\frac{dy}{dx} = 0$ , we reach the minimum.
  - We can use Gradient Descent Algorithm: (Y<sup>output</sup> – Y)<sup>2</sup> is a function of w

For each w, we want to find a value to make the function value smallest.  $(Y^{output} - Y)^2$ 



Step:

- 1. Find  $\frac{dy}{dx}$  of  $(Y^{output} Y)^2$  at a point A.
- 2. If  $\frac{dy}{dx}$  result is negative, we choose another point B to get a result.
- 3. Repeat this process to let the  $\frac{dy}{dx}$  closer to 0
- 4. Until very close to 0 (maybe not 0)
- 5. Then it is the minimum we find

- 4. To get the formula
  - Initialize  $W_h$ ,  $W_w$ , and  $W_0$  (Random values)
  - Calculate the output Y<sup>output</sup>
  - Update weights
    - $w_i = w_i + \Delta w_i$
    - $\Delta w_i = 2 * \alpha (Y Y^{output}) \frac{\partial Y^{output}}{\partial w_i}$
    - $\alpha$  is a small constant
  - Repeat the above steps until no more update

#### **B.** From Logistic Regression to Neural Networks

### a) The simplest neural networks

One input node and one output nodes



#### b) From logistic regression to neural networks

Several input nodes and several output nodes

We can create function of functions (such as layer L<sub>3</sub> below)



<u>Advantage:</u> Fast prediction; Successful in real-life problems; High tolerance to noisy data <u>Disadvantage:</u> Long training time; Poor interpretability

# c) From neural networks to deep learning

Successful deep learning application: AlphaFold



# **III. Self-learning Materials**

Introduction to data mining: Chapter 5.4 & Appendix E

- Problems of neural network
- Decision tree/SVM/Bayesian...
- Model overfitting
- Cross-validation (next lecture)