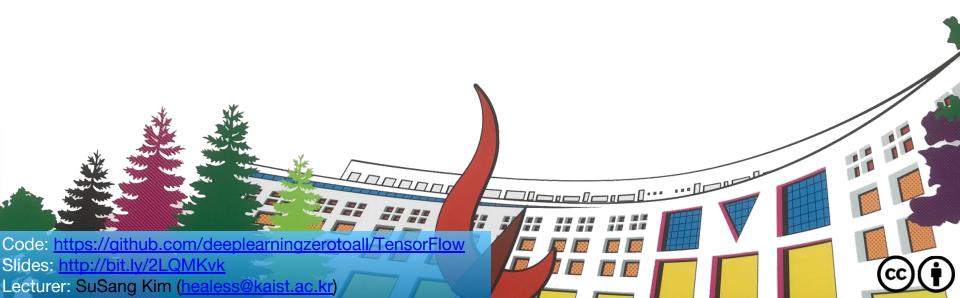
# ML/DL for Everyone Season2



Lab 05-1 Logistic Regression



## **Logistic Regression**

- What is Logistic Regression?
  - Classification
  - Logistic vs Linear
- How to solve?
  - Hypothesis Representation
  - Sigmoid/Logistic Function
  - Decision Boundary
  - Cost Function
  - Optimizer (Gradient Descent)
- Codes
- Summary

### Classification

### What is Binary(Multi-class) Classification?

variable is either 0 or 1 (0:positive / 1:negative)

- Exam : Pass or Fail
- Spam : Not Spam or Spam
- Face : Real or Fake
- Tumor : Not Malignant or Malignant

To start with machine learning, you must encode variable [0,1]

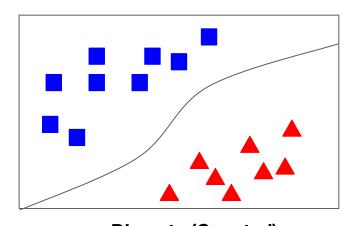
```
[Python Code]
```

```
x_train = [[1, 2], [2, 3], [3, 1], [4, 3], [5, 3], [6, 2]]
y_train = [[0], [0], [0], [1], [1], [1]] # One Hot
```

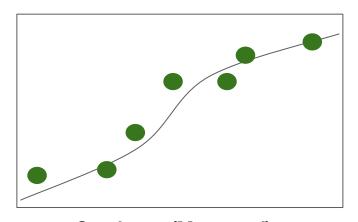
### Logistic vs Linear

### What is the difference between logistic and linear?

VS



Discrete (Counted)
Shoe Size /The number of workers in a company



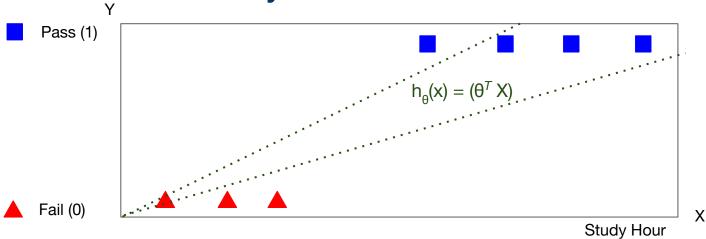
Continous (Measured)
Time / Weight / Height

#### [Python Code]

```
Logistic_Y= [[0], [0], [0], [1], [1], [1]] # One Hot
Linear_Y = [828.659973, 833.450012, 819.23999, 828.349976, 831.659973] # Numeric
```

## **Hypothesis Representation**

Study Hours ⇔ Pass/Fail



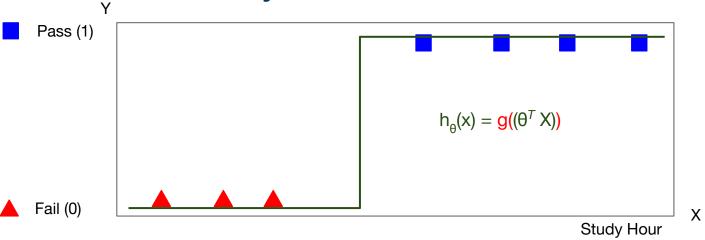
When using linear regression we did  $h_{\theta}(x) = (\theta^T X)$ But start with binary classification, Y is only 0 or 1 So we need new function.

#### [Tensorflow Code]

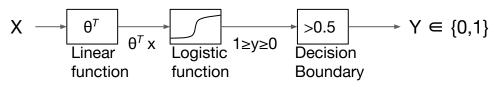
hypothesis = tf.matmul(X,  $\theta$ ) + b #linear  $\theta^T$  is an [1xn+1] matrix /  $\theta_i$  are parameters

## **Hypothesis Representation**

Study Hours ⇔ Pass/Fail

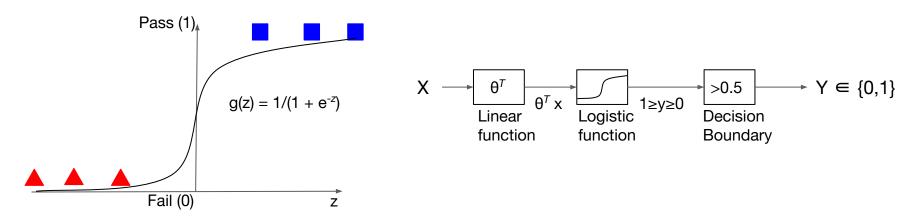


For classification hypothesis representation we do  $h_{\theta}(x) = g((\theta^T X))$ What is g(z) Function? (linear function :  $z = \theta^T X$ )



## Sigmoid (Logistic) function

### g(z) function out value is between 0 and 1



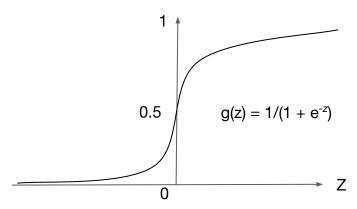
Where we define  $g(z) \rightarrow z$  is a real number  $\rightarrow g(z) = e^{z}/(e^{z} + 1) = 1/(1 + e^{-z})$ 

#### [Tensorflow Code]

```
hypothesis = tf.sigmoid(z) # z= tf.matmul(X, W) + b : \theta=W hypothesis = tf.div(1., 1. + tf.exp(z))
```

## **Decision Boundary**

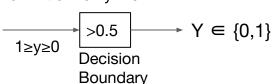
### Linear / Non-linear decision boundary

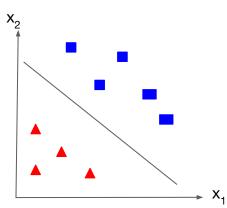


if z is positive, g(z) is greater than 0.5

$$\theta^T x >= 0$$
 then  $y = 1$ 

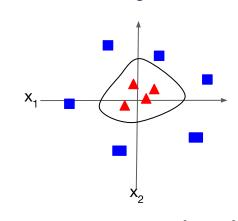
$$\theta^T x < 0$$
 then  $y = 0$ 





$$\begin{aligned} &h_{\theta}(x) = g(\theta_0 + \theta_1 x_1 + \theta_2 x_2) \\ &\theta^T \text{ is a row vector} = [-3,1,1] \\ &-3 + x_1 + x_2 >= 0 \\ &\textbf{Decision Boundary } x_1 + x_2 = 0 \end{aligned}$$

 $\Re \theta^{T}$  is an [1 x n+1] matrix /  $\theta_{i}$  are parameters / (h = hypothesis)



$$h_{\theta}(x) = g(\theta_0 + \theta_1 x_1 + \theta_3 x_1^2 + \theta_4 x_2^2)$$

$$\theta^{T} \text{ was } [-1,0,0,1,1]$$

$$-1 + x_1^2 + x_2^2 >= 0$$
Decision Boundary  $x_1^2 + x_2^2 = 1$ 

#### [Tensorflow Code]

predicted = tf.cast(hypothesis > 0.5, dtype=tf.int32)