

# Logistic Regression

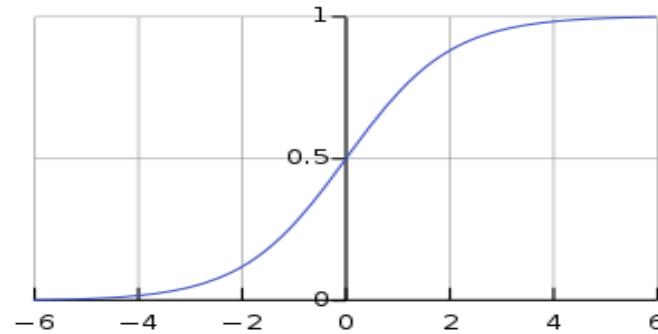
# Logistic Regression

- This algorithm is used for classification type problems
- Types of Logistic Regression:
  - Binary
  - Multinomial
  - Ordinal
- We are going to cover Binary & Multinomial Logistic Regression

# Logistic Response Function

- Standard logistic function on 2-dimensional plane is given by the following expression given on the right.
- From the graph, it is evident that the value of the  $f(x)$  ranges between 0 and 1.
- This function is also called sigmoid function and has a wide usage in various other algorithms such as neural network.

$$y = f(x) = \frac{1}{1 + e^{-x}}$$



# Logistic Response Function

- The same function in the m-dimensional space can be written in the following way:

$$y = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_m x_m)}}$$

Where

$\beta_0, \beta_1, \beta_2, \dots, \beta_m$ : Coefficients of the variables in m-dimensional space

- For any values of  $\beta_0, \beta_1, \beta_2, \dots, \beta_m$  and  $x_1, x_2, \dots, x_m$ , the value of  $y$  always between 0 and 1.
- We can denote  $y$  by probability  $p$ .

# Odds

$$p = \frac{1}{1+e^{-(\beta_0+\beta_1x_1+\beta_2x_2+\dots+\beta_mx_m)}}$$

$$1 - p = \frac{e^{-(\beta_0+\beta_1x_1+\beta_2x_2+\dots+\beta_mx_m)}}{1 + e^{-(\beta_0+\beta_1x_1+\beta_2x_2+\dots+\beta_mx_m)}}$$

$$\frac{p}{1-p} = e^{(\beta_0+\beta_1x_1+\beta_2x_2+\dots+\beta_mx_m)}$$

- The ratio  $\frac{p}{1-p}$  is called odds. For any values of  $\beta_0, \beta_1, \beta_2, \dots, \beta_m$  and  $x_1, x_2, \dots, x_m$ , odds always ranges from 0 to  $\infty$ .

# Interpreting Logistic Function

- In our binary classification, let us consider 0 and 1 as two possible outcomes, with 0 as non-occurrence of a particular event and 1 as occurrence of the particular event.
- $p$  in our expression, is considered as probability of occurrence of the event and  $1 - p$  as non-occurrence of the event
- Hence, the ratio  $\frac{p}{1-p}$  is ratio of probability of occurrence to the probability of non-occurrence of the event.

# Logit Function

$$\frac{p}{1-p} = e^{(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_m x_m)}$$

$$\log(\text{odds}) = \log\left(\frac{p}{1-p}\right) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_m x_m$$

- The ratio  $\log\left(\frac{p}{1-p}\right)$  is called logit function. For any values of  $\beta_0, \beta_1, \beta_2, \dots, \beta_m$  and  $x_1, x_2, \dots, x_m$ ,  $\log(\text{odds})$  always range from  $-\infty$  to  $\infty$ .

# Parameter Calculation

- Parameters  $\beta_0, \beta_1, \beta_2, \dots \beta_m$  are calculated with the help of maximum likelihood method.
- In Python, we make use of the function `LogisticRegression()` from `sklearn.linear_model`



The background of the slide features a subtle, abstract graphic composed of numerous small, light gray and white geometric shapes, primarily triangles and squares, arranged in a pattern that suggests depth and movement, resembling a stylized forest or a complex data visualization.

Questions?