

Day-10, Oct-12, 2024 (Dashami-DashainTika 2081BS)

Logistic Regression Mathematics Glimpse

logistic Regression is used to predict the outcome of input in the range of $[0, 1]$ probabilities

→ used to classify either outcome is 0 or 1

→ Spam calls or mail, diseased with tuberculosis or not,

wine quality good or bad

$$p(x) = \frac{P}{1+P}$$

where $p(x)$ is sometimes 0 called Odds.

So,

$$y = \beta_0 + \beta_1 x$$

where,

$\beta_0 = y$ -intercept

$\beta_1 \Rightarrow$ Coefficient of x or slope

$x \Rightarrow$ input.

How to find the Coefficient of x and β_0 ?

① Linear Combination =

$$z = \beta_0 + \beta_1 x$$

② Predicted probabilities : $p = \frac{1}{1 + e^{-z}}$

③ Gradient Calculations: (Gradient β_0 and Gradient β_1):
for β_0 : $\frac{1}{n} \sum_{i=1}^n (p_i - y_i)$

for β_1 : $\frac{1}{n} \sum_{i=1}^n (p_i - y_i) x_i$

Where x can be input like
hours, Speed of note,
wine quality, diabetes
level

● Coefficient updates:
for β_0 : $\beta_0 - \alpha \cdot \text{Gradient } \beta_0$
for β_1 : $\beta_1 - \alpha \cdot \text{Gradient } \beta_1$

Example:

① Initialize Parameters

- Set up β_0 and β_1 (0, 0)
- Define rate of learning α , threshold and iterations
(only for programming & coding part)

② Logistic Regression:

$$p = \frac{1}{1 + e^{-z}}$$

$$\text{Where } z = \beta_0 + \beta_1 x$$

and we compute p for each each data points

⑧ Calculate the Gradients:

for: β_0

$$\text{gradient}_0 \Rightarrow \frac{1}{n} \sum (p_i - y_i)$$

Here p is calculated below

$$= \frac{1}{2} \sum ([0.5, 0.5] - [0, 1])$$

$$= \frac{1}{2} \sum ([0.5, -0.5])$$

$$= \frac{1}{2} [0.5 + (-0.5)]$$

$$= \frac{1}{2} \times 0 \Rightarrow 0$$

for β_1 :

$$\text{Gradient}_1 \Rightarrow \frac{1}{n} \sum (p_i - y_i) \cdot x_i$$

So, before that

$$z = \beta_0 + \beta_1 \cdot x$$

$$\Rightarrow 0 + 0 \cdot [x]$$

$$z = [0 \ 0]$$

$$[\beta_0 = 0, \beta_1 = 0 \text{ Assume}]$$

$$\text{So, } p = \frac{1}{1 + e^{-z}}$$

$$\Rightarrow \frac{1}{1 + e^{-0}}$$

$$[p = \frac{1}{2} \Rightarrow [0.5, 0.5]]$$

$$g_{\text{gradient}_1} \Rightarrow \frac{1}{2} \sum (p - \text{pass}(Y_i)) \cdot x$$

$$\begin{array}{l} x = 1, 2 \\ y = 0, 0 \end{array}$$

$$= \frac{1}{2} \sum ([0.5, 0.5] - [0, 1]) \cdot [1, 2]$$

$$\Rightarrow \frac{1}{2} \sum ([0.5, -0.5]) \times [1, 2]$$

$$\Rightarrow \frac{1}{2} \sum ([0.5, -1])$$

$$\Rightarrow \frac{1}{2} \times (-0.5) \quad \Rightarrow -0.25$$

Update Gradients for β_0 and β_1 so

$$\text{Old } \beta_0, \beta_1 \Rightarrow [0, 0] \quad \text{But New } \beta_0, \beta_1 \Rightarrow [0, -0.25]$$

Continue the process until convergence
which means β_0 and β_1 have very negligible
change example

$$\left[\begin{array}{l} \beta_0 \Rightarrow 0.5678 \\ \beta_0 \Rightarrow 0.5675 \end{array} \right]$$

So, the overall steps is

- ① Calculate P
- ② Gradient Calculate β_0, β_1
- ③ Update Gradient [if converges Stop else Continue]