

Cost V3.
for MSE loss,

$$J_{\text{theta}} = \frac{1}{N} \sum_{i=1}^N (h_{\theta}(x_i) - y_i)^2$$

In case of logistic regression

$$h_{\theta}(x) = \frac{1}{1 + e^{-\theta x}}$$

$$\frac{\partial J_{\theta}}{\partial \theta} = \frac{2 \sum_{i=1}^N [h_{\theta}(x_i) - y_i] \left(\frac{\partial h_{\theta}(x_i)}{\partial \theta} \right)}{N}$$

for 1 dataset,

$$= 2 [h_{\theta}(x) - y] \frac{\partial h_{\theta}(x)}{\partial \theta}$$

$$= 2 [h_{\theta}(x) - y] \times \frac{1 - h_{\theta}(x)}{(1 + e^{-\theta x})^2} \times -x e^{-\theta x}$$

$$= 2 [h_{\theta}(x) - y] \times \frac{1 + e^{-\theta x} - 1}{(1 + e^{-\theta x})^2} \times x$$

$$= 2 [h_{\theta}(x) - y] \times \frac{1}{e^{\theta x} + 1} \times \left[1 - \frac{1}{e^{\theta x} + 1} \right] x$$

$$= 2[h_0(m) - y] \times h_0(m) [1 - h_0(m)] \ln$$

Hence, when

$$h_0(m) = 1, \frac{\partial J_0}{\partial \theta} = 0$$

$$h_0(m) = 0, \frac{\partial J_0}{\partial \theta} = 0$$

Here gradient will be same everytime as its derivative will always be 0, so model will not be able to learn.

Using cross entropy loss,

$$\frac{\partial J_0}{\partial \theta} = [h_0(m) - y] \ln$$

Using this model will be ~~not~~ able to learn as its value is ~~not~~ is not constant like before.

Ans. $Y = X\beta + \epsilon$

$$Y - \epsilon = X\beta$$

$$\Rightarrow \hat{Y} = X\beta$$

$$J(\beta) = \frac{1}{N} \sum_{i=1}^N |y_i - \hat{y}_i|^2$$

$$= \frac{1}{N} \sum_{i=1}^N |X\beta - y|^2$$

$$\frac{\partial J}{\partial \beta} = \frac{\partial}{\partial \beta} \sum_{i=1}^N (X\beta - y)^2$$

For datapoint

$$\frac{\partial J}{\partial \beta} = \frac{\partial [X\beta - y]^2}{\partial \beta}$$

$$= \frac{\partial [X\beta - y]^T [X\beta - y]}{\partial \beta}$$

~~$$= \frac{\partial [X^T X \beta - X^T y]}{\partial \beta}$$~~

Using pt rule,

~~$$X^T X \beta = X^T y + X^T X \beta - X^T y$$~~

$$= \frac{\partial [X^T X \beta - (X^T y) - y^T (X\beta) + y^T y]}{\partial \beta}$$

$$= \frac{1}{2} \left[X^T B^T X B - 2 X^T B^T y - y^T y \right]$$

$$= \frac{1}{2} \left[X^T B^T X B - 2 X^T B^T y \right]$$

$$= X^T X \frac{\partial B^T B}{\partial B}$$

$$= 2 X^T X B - 2 X^T y$$

$$\text{for } \frac{\partial J_0}{\partial B} = 0;$$

$$2 X^T X B = 2 X^T y$$

$$\Rightarrow X^T X B = X^T y$$

$$\Rightarrow (X^T X)^{-1} (X^T X) B = (X^T X)^{-1} X^T y$$

$$\Rightarrow B = (X^T X)^{-1} X^T y$$

Solution will exist ~~only~~ when $(X^T X)$ have inverse exist.
 so, there must not be a duplicate data.