

Maximum Likelihood Estimator Derivation

This is the procedure to compute the **Estimates** for the model **Coefficients** .

Continuing from [Sigmoid Function](#) Derivation Section we got :

$$P(X_i) = \frac{1}{1 + e^S}$$

Let $S = -\vec{X}\vec{\beta}$

Finding the **Likelihood function** for the coefficients β

$$L(\vec{\beta}) = \Pr(Y_1, Y_2 \dots, Y_n)$$

$$L(\vec{\beta}) = \prod_{i=1}^n \Pr(Y_i = 1)$$

$$L(\vec{\beta}) = \prod_{i=1}^n P(\vec{X}_i)^{Y_i} (1 - P(\vec{X}_i))^{1-Y_i}$$

$$\log(L(\vec{\beta})) = \log \left(\prod_{i=1}^n P(\vec{X}_i)^{Y_i} (1 - P(\vec{X}_i))^{1-Y_i} \right)$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n Y_i \log(P(\vec{X})) + (1 - Y_i) \log(1 - P(\vec{X}))$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n Y_i \log\left(\frac{1}{1 + e^S}\right) + (1 - Y_i) \log\left(\frac{e^S}{1 + e^S}\right)$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n -Y_i \log(1 + e^S) + (1 - Y_i)(\log(e^S) - \log(1 + e^S))$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n -Y_i \log(1 + e^S) + \log(e^S) - \log(1 + e^S) - Y_i \log(e^S) + Y_i \log(1 + e^S)$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n \log(e^S) - \log(1 + e^S) - Y_i \log(e^S)$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n \log(e^S)(1 - Y_i) - \log(1 + e^S)$$

$$\log(L(\vec{\beta})) = \sum_{i=1}^n S(1 - Y_i) - \log(1 + e^S)$$

Derivation with respect to β

$$\log(L(\vec{\beta})) = \sum_{i=1}^n (-\vec{X}\vec{\beta})(1 - Y_i) - \log(1 + e^{-\vec{X}\vec{\beta}})$$

$$\frac{d \log(L(\vec{\beta}))}{d \vec{\beta}} = \sum_{i=1}^n -\vec{X} + \vec{X} Y_i - \frac{1}{1 + e^{\vec{X}\vec{\beta}}} (-\vec{X} e^{-\vec{X}\vec{\beta}})$$

$$\frac{d \log(L(\vec{\beta}))}{d \vec{\beta}} = \sum_{i=1}^n -\vec{X}_i (1 - Y_i) + \frac{e^{-\vec{X}\vec{\beta}}}{1 + e^{\vec{X}\vec{\beta}}} \vec{X}_i$$

$$\frac{d \log(L(\vec{\beta}))}{d \vec{\beta}} = \sum_{i=1}^n \vec{X}_i \left[-(1 - Y_i) + \frac{e^{-\vec{X}\vec{\beta}}}{1 + e^{\vec{X}\vec{\beta}}} \right]$$

Now we solve for β

$$\sum_{i=1}^n \vec{X}_i \left[-(1 - Y_i) + \frac{e^{-\vec{X}\vec{\beta}}}{1 + e^{\vec{X}\vec{\beta}}} \right] = 0$$

- Unlike Linear Regression, the **Likelihood function** in the logistic regression is **nonlinear** in the parameters β , the gradient of the log-likelihood do not yield a closed-form solution for β
- We use **iterative numerical optimization methods**