$$y = f(w^Tx + w_0)$$

$$P(X \mid C_{K}) = \frac{1}{(2\pi)^{d|2} |\Sigma|^{1/2}} e^{xP} \left\{ -\frac{1}{2} (x - \mu_{K})^{T} \Sigma^{-1} (x - \mu_{K}) \right\}$$

$$P(C_1|x) = \frac{P(x|C_1)P(C_1)}{P(x|C_1)P(C_1) + P(x|C_2)P(C_2)}$$

$$\frac{1 + P(x | C_2)P(C_2)}{P(x | C_1)P(C_1)}$$

$$= \frac{1}{1 + e^{-\ln P(x_1C_1)P(C_1)}}$$

$$= \frac{1}{-Z}$$

$$Z = \ln \frac{P(x|c_i)P(c_i)}{P(x|c_i)P(c_i)} = \omega^T x + w_0$$

STEP 1 Read data file (pandas) STEP 2 Process data file (i) Drop column id (ii) map label column to 0 and 1 STEP 3 Split data frame into train, Val of Test STEP @ Normalize the dataset STEP & Initialize weights and biases & learning Pate STEP @ for epoch in large (10000) $z = \mathbf{\Theta}^{\mathsf{T}} \times + \mathbf{b}$ a = T(Z) $L = -\frac{(y \log a + (1-y) \log (1-a))}{m}$ $= -\frac{1}{m} \left(y \log \alpha + (1-y) \log (1-\alpha) \right)$ $= -\frac{1}{m} \left(y \log \left(\overline{\tau(z)} \right) + (1-y) \log \left(1 - \overline{\tau(z)} \right) \right)$

Logistic Regression - Method 1

$$\begin{array}{l}
\omega = \omega - \eta \Delta O \\
b = b - \eta \Delta b
\end{array}$$

$$L = -\frac{1}{m} \left\{ y \log \alpha + (1-y) \log (1-\alpha) \right\}$$

$$= -\frac{1}{m} \left\{ y \log \sigma(z) + (1-y) \log (1-\sigma(z)) \right\}$$

$$= -\frac{1}{m} \left\{ y \log \sigma(z) + (1-y) \log (1-\sigma(z)) \right\}$$

$$= -\frac{1}{m} \left\{ y \cdot \frac{1}{\sigma(z)} \cdot \frac{\partial}{\partial \theta_1} \sigma(z) + (1-y) \log (1-\sigma(z)) \right\}$$

$$= -\frac{1}{m} \left\{ y \cdot \frac{1}{\sigma(z)} \cdot \frac{\partial}{\partial \theta_1} \sigma(z) + (1-y) \cdot \frac{1}{1-\sigma(z)} \frac{\partial}{\partial \theta_1} (1-\sigma(z)) \right\}$$

$$= -\frac{1}{m} \left\{ y \cdot \frac{1}{\sigma(z)} \sigma(z) \left(1-\sigma(z) \right) \frac{\partial}{\partial \theta_1} \left(-z \right) \right\}$$

$$= -\frac{1}{m} \left\{ y \cdot (1-\sigma(z)) \times_1 + (1-y) \sigma(z) \left(-x_1 \right) \right\}$$

$$= -\frac{1}{m} \left\{ y - y \sigma(z) - \sigma(z) + y \sigma(z) \right\} \times_1$$

$$\Delta \theta_1 = -\frac{1}{m} \left\{ y - y \sigma(z) \right\} - \sigma(z) + y \sigma(z) \right\}$$

For updating Bias,
$$\Delta b = -\frac{1}{m} \left\{ y - \sigma(z) \right\}. 1$$

