

Ans 2

We get the best accuracy when we take n components = 15 because that way we have used the maximum number of features in this case for analysis.

Ans 5 (a)

$$d_i = y_i (\beta^T x_i + \beta_0)$$

Changing activation function does not affect the update rule.

→ Distance does not depend on the activation function.

$$\beta_{\text{new}} \Rightarrow \beta - \eta \frac{\partial d_i}{\partial \beta}$$

$$\beta_{0\text{new}} \Rightarrow \beta_0 - \eta \frac{\partial d_i}{\partial \beta_0}$$

$$(b) \quad \phi(\beta, \beta_0) = - \sum_{i=1}^N y_i (\beta^T x_i + \beta_0)$$

We know that $\beta^T \beta = 1$

$$g(\beta) = \beta^T \beta - 1$$

$$\alpha(\beta, \beta_0) = \phi(\beta, \beta_0) + \lambda(g(\beta))$$

$$\alpha(\beta, \beta_0) = - \sum_{i=1}^N y_i (\beta^T x_i + \beta_0) + \lambda(\beta^T \beta + 1)$$

$$\frac{\partial L}{\partial \beta} = - \sum_{i=1}^N y_i (x_i) + \lambda(2\beta) = 0$$

$$\sum_{i=1}^n x_i y_i + 2\lambda \beta = 0$$

$$\begin{aligned} \frac{\partial \alpha}{\partial \beta_0} &= - \sum_{i=1}^n y_i = 0 \\ &= \sum_{i=1}^n y_i = 0 \end{aligned}$$

Ans 6

$$y_1 = \sigma(\beta_{11} x + \beta_{01})$$

$$y_2 = \text{sig}(\beta_{12} y_1 + \beta_{02})$$

$$x \longrightarrow \bigcirc \xrightarrow{\uparrow y_1} \bigcirc y_2$$

For misclassified samples

$$d_i = -y_i (\beta_{12} y_{1i} + \beta_{02})$$

$$\frac{\partial d_i}{\partial \beta_{11}} = -y_i \left(\beta_{12} \frac{\partial y_{1i}}{\partial \beta_{11}} \right)$$

$$\sigma'(x) = \sigma(x) [1 - \sigma(x)]$$

$$= -y_i (\beta_{12} \sigma(\beta_{11} x_i + \beta_{01}) x_i)$$

$$= -y_i (\beta_{12} [\sigma(\beta_{11} x_i + \beta_{01}) (1 - \sigma(\beta_{11} x_i + \beta_{01}))] x_i)$$

$$\frac{-y_i x_i \beta_{12} e^{-v_i}}{(1 + e^{-v_i})^2}$$

Now,

$$\frac{\partial d_i}{\partial \beta_{11}} = -y_i \left(\beta_{12} \frac{\partial y_{1i}}{\partial \beta_{01}} \right)$$

$$\frac{\partial y_{1i}}{\partial \beta_{01}} = \sigma'(\beta_{11} x_i + \beta_{01})$$

$$= \frac{e^{-v_i}}{(1 + e^{-v_i})^2}$$

$$\frac{\partial d_i}{\partial \beta_{01}} = -y_i \frac{\beta_{12} e^{-v_i}}{(1 + e^{-v_i})^2}$$

$$\frac{\partial d_i}{\partial \beta_{02}} = -y_i (y_{1i})$$

$$= -y_i [\sigma(\beta_{11} x + \beta_{01})]$$

$$\frac{\partial d_i}{\partial \beta_{02}} = -y_i$$

$$\beta_{01} \rightarrow \beta_{01} - \eta \frac{\partial d_i}{\partial \beta_{01}}$$

$$\beta_{12} \rightarrow \beta_{12} - \eta \frac{\partial d_i}{\partial \beta_{12}}$$

$$\beta_{02} \rightarrow \beta_{02} - \eta \frac{\partial d_i}{\partial \beta_{02}}$$

$$\beta_{11} \rightarrow \beta_{11} - \eta \frac{\partial d_i}{\partial \beta_{11}}$$