

Question 5@ Lecture 16

to prove :- if an activation function is sign function if we replace it with sigmoid function; we need to prove update parameters don't change

distance eqn. is as follows.

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

a pt.  $x_i$  with true label is  $y_i$   
prediction of perceptron

$$\beta^T x_i + \beta_0 \geq 0 + 1$$

$$< 0 - 1$$

changing activation fn will not influence update parameters because distance is not dependent on the activation fn.

$$\beta^{\text{new}} \leftarrow \beta_{\text{old}} - \eta \frac{\partial \mathcal{L}}{\partial \beta}$$

$$= \beta_{\text{old}} - \eta (-y_i x_i)$$

$$\beta_0^{\text{new}} \leftarrow \beta_0^{\text{old}} - \eta (-y_i)$$

⑤ Given :

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

$$\min \phi(\beta, \beta_0) \text{ s.t. } \|\beta\| = 1$$

$$\beta^T \beta = 1$$

$$\mathcal{L}(\beta, \beta_0, \lambda) = \phi(\beta, \beta_0) - \lambda g(\beta)$$

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

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

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

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

$$\frac{\partial \mathcal{L}}{\partial \beta_0} = -\sum_{i=1}^N y_i = 0$$

$$\frac{\partial \mathcal{L}}{\partial \lambda} = -(\beta^T \beta - 1) = 0$$

$$= \beta^T \beta = 1$$



$$\beta_{new} \leftarrow \beta - \eta \frac{\partial L}{\partial \beta}$$

$$\beta_{new} \leftarrow \beta_0 - \eta \frac{\partial L}{\partial \beta_0}$$

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

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

$$d_i^2 = -y_i (\beta_{12} y_{1i} + \beta_{02})$$

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

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

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

$$\beta_{11} = \eta \frac{\sum d_i}{\sum \frac{\partial d_i}{\partial \beta_{11}}} \rightarrow \beta_{11}$$

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

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

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

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

$$\frac{\partial d_i}{\partial \beta_{12}} = -y_i (y_{11'})$$

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

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

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

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

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