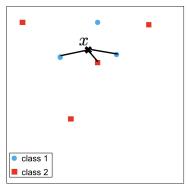
I.) X와 가장 가까운 요소 3가지를 보자 (∵k=3)



■ ● 순으로 가까운 요소이므로,

3개중 2개를 차지하는 ● 에 의해 X는 class I으로 estimated 된다.

$$x = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}, \overline{W} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}, b = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

first element of s represents the score for "cat" so the score for "cat" is 3

second element of s represents the score for "dog" so the score for "dog" is 6

third element of s represents the score for "ship" so the score for "ship" is Γ

3)
$$L = \sum_{j \neq y_i} (M \text{ and } (0, S_j - S_{y_i} + I))$$

$$\begin{bmatrix} s_1 = \begin{bmatrix} 3 \\ 5 \\ -1 \end{bmatrix}, \quad s_2 = \begin{bmatrix} 1 \\ 4 \\ 2 \end{bmatrix}, \quad s_3 = \begin{bmatrix} 2 \\ 2 \\ -3 \end{bmatrix}$$

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Į		Predicted class: Positive	Predicted class: Negative
_[actual class: Positive	85 2 T P	15 = F N
	actual class: Negative	890 ZFP	10 = TN

$$accuracy = \frac{TP+TN}{TP+FP+FN+TN}$$

= 0.095

precision =
$$\frac{TP}{TP + FP}$$

$$= \frac{85}{85 + 890}$$

$$= 0.0872$$

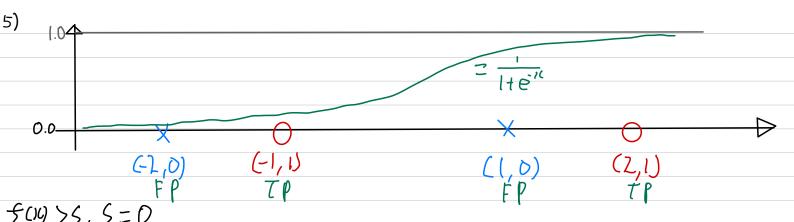
$$recall = \frac{TP}{TP + FN}$$

$$= \frac{85}{85 + 15}$$

$$0.85$$

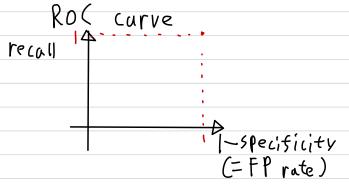
$$F-score = \frac{2 \times precision \times recall}{precision + recall}$$

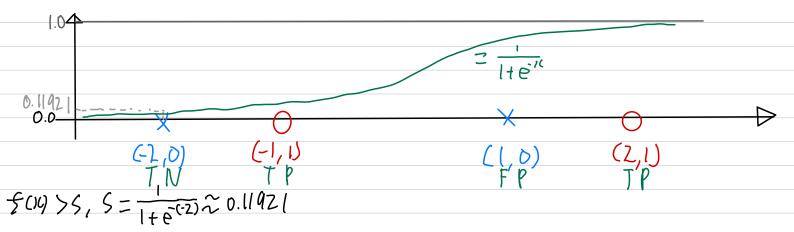
$$\frac{2 \times 0.0872 \times 0.85}{0.0872 + 0.85}$$
$$= 0.1582$$



$$recall = \frac{TP}{TP+FN} = \frac{2}{2+0} = 1$$

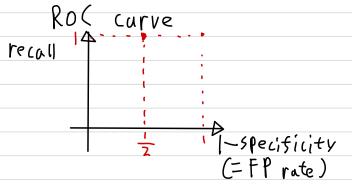
$$FP rate = \frac{FP}{TN+FP} = \frac{2}{0+2} = |$$

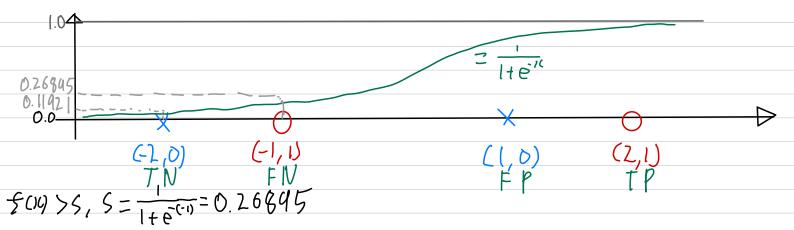




$$recall = \frac{TP}{TP + FN} = \frac{2}{2 + 0} = 1$$

$$FP rate = \frac{FP}{TN+FP} = \frac{1}{1+1} = \frac{1}{2}$$

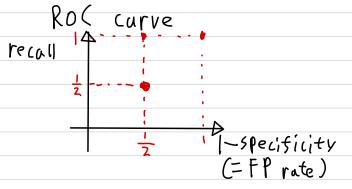


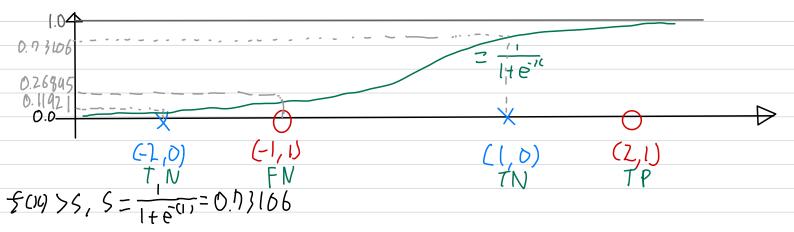


$$I = TP$$
 $I = FP$

recall =
$$\frac{TP}{TP+FN} = \frac{1}{1+1} = \frac{1}{2}$$

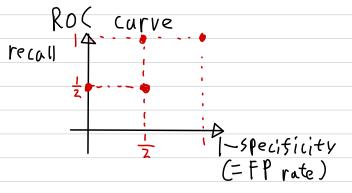
$$FP \text{ rate} = \frac{FP}{TN+FP} = \frac{1}{1+1} = \frac{1}{2}$$

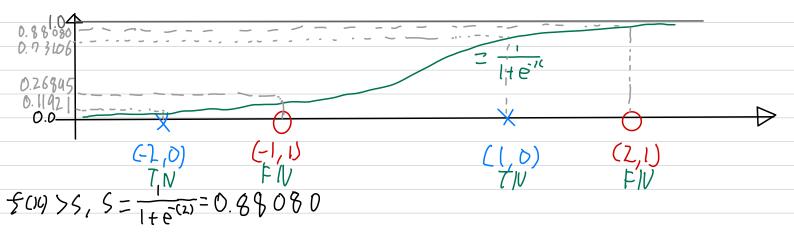




$$recall = \frac{TP}{TP + FN} = \frac{b}{1 + 0} = 0$$

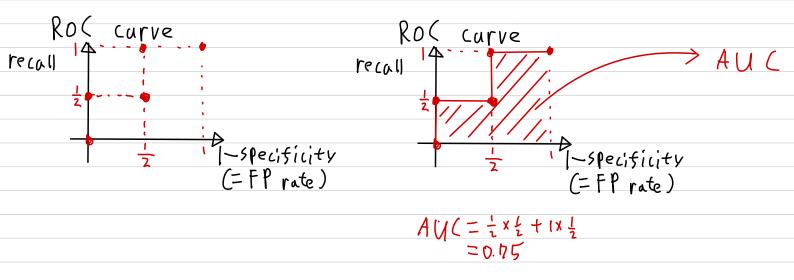
$$FP rate = \frac{FP}{TN+FP} = \frac{1}{1+1} = \frac{1}{2}$$





$$recall = \frac{TP}{TP + FN} = \frac{6}{100} = 0$$

$$FP \text{ rate} = \frac{FP}{TN+FP} = \frac{O}{O+2} = 0$$



6)
$$\operatorname{softmax}(x) = \frac{\exp(x_{j})}{\sum_{j=1}^{n} \exp(x_{j})} (j=1,2,...,n)$$

$$s = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} \quad \begin{array}{c} \boxed{\frac{e^{\flat}}{e^{\flat} + e^{\flat} + e^{\flat}}} \\ \boxed{\frac{e^{\flat}}{e^{\flat} + e^{\flat} + e^{\flat}}} \\ \boxed{\frac{e^{\flat}}{e^{\flat} + e^{\flat} + e^{\flat}}} \\ \boxed{\frac{e^{\flat}}{e^{\flat} + e^{\flat} + e^{\flat}}} \end{array} \quad \begin{array}{c} 0.665241 \\ 0.244728 \\ 0.090031 \\ \boxed{\frac{e^{\flat}}{e^{\flat} + e^{\flat} + e^{\flat}}} \end{array}$$

$$s = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

ground truth label for s is class 2

$$L = -\log(0.244728)$$

$$\approx 1.407608$$

$$y = softmax (W_2 ReLU(W_1 x))$$

$$W_1 = \begin{bmatrix} -2 & 1 \\ 3 & -4 \end{bmatrix}, \quad W_2 = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}, \quad \chi = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$W_1 \cdot \chi = \begin{bmatrix} -2 & 1 \\ 3 & -4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ -q \end{bmatrix}$$

ReLU(W,x) =
$$\left(\frac{1}{2}\right)\left(\frac{1}{2}\right)\left(\frac{1}{2}\right)\left(\frac{1}{2}\right)$$

$$W_2$$
 ReLU($W_1 \times$) = $\begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix}$ $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ = $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$

$$y = softmax (W_2 ReLU(W_1 x)) = \begin{bmatrix} \frac{e^2}{e^2 + e^4} \\ \frac{e^4}{e^2 + e^4} \end{bmatrix} = \begin{bmatrix} 0.7311 \\ 0.2689 \end{bmatrix}$$

9)

gradient descent를 사용하면,

$$(\chi, \gamma) \rightarrow (\chi - \chi \frac{\sigma L(\chi, \gamma)}{\sigma \chi}, \gamma - \chi \frac{\sigma L(\chi, \gamma)}{\sigma \gamma})$$

$$= (\chi - \chi \frac{\sigma L(\chi + 3 \chi \gamma)}{\sigma \chi}, \gamma - \chi \frac{\sigma (2 \chi + 3 \chi \gamma)}{\sigma \gamma})$$

$$= (\chi - \chi (2 + 3 \gamma), \gamma - \chi (3 \chi))$$

$$\chi := \chi - 2\alpha - 3\alpha \gamma$$

 $\gamma := \gamma - 3\alpha \chi$