

Q1

문제 1-1)

$$A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 3 \\ 0 & 0 & 3 \end{bmatrix} \quad |A| = 1 \times (2 \times 3 - 0 \times 3) + 0 \times (0 \times 3 - 3 \times 0) + 1 \times (0 \times 0 - 0 \times 2) = 6$$

문제 1-2)

$$A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 3 \\ 0 & 0 & 3 \end{bmatrix} \quad I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad A \cdot x = x \lambda$$

$$\Leftrightarrow x(A - I\lambda) = 0$$

$$\therefore \det(A - I\lambda) = 0$$

$$\Leftrightarrow A - I\lambda = \begin{bmatrix} 1-\lambda & 0 & 1 \\ 0 & 2-\lambda & 3 \\ 0 & 0 & 3-\lambda \end{bmatrix}$$

$$\Rightarrow (1-\lambda)((2-\lambda)(3-\lambda) - 3 \times 0) + 0 \times (0 \times (3-\lambda) - 1 \times 0) + 1 \times (0 \times 0 - 0 \times (2-\lambda))$$

$$= -(\lambda-1)(\lambda-2)(\lambda-3) = 0$$

$$\therefore \lambda = 1, 2, 3$$

(case-1) $\lambda = 1$

$$(A - \lambda I) \cdot x = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = 0 + b + 2c = 0 \quad \therefore b = -2c \rightarrow \vec{x} = (x, -2t, t)$$

(case-2)

$$(A - \lambda I) \cdot x = \begin{bmatrix} -1 & 0 & 1 \\ 0 & 0 & 3 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = -a + c = 0 \quad \therefore a = c \rightarrow \vec{x} = (t, y, t)$$

(case-3)

$$(A - \lambda I) \cdot x = \begin{bmatrix} -2 & 0 & 1 \\ 0 & -1 & 3 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = -2a - b = 0 \quad \therefore b = -2a \quad \therefore \vec{x} = (-2t, t, z)$$

Q2

$$B = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 3 & 0 & 1 \end{bmatrix} \quad B = P^{-1}DP$$

$$i) \det(B - \lambda I) = \begin{vmatrix} -\lambda & 0 & 0 \\ 0 & -\lambda & 0 \\ 3 & 0 & 1-\lambda \end{vmatrix} = (-\lambda)(-\lambda)(1-\lambda) = 0 \quad \therefore \lambda = 0 \text{ or } 1$$

$$\therefore b = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

① $\lambda = 0$ ② $\lambda = 1$

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 3 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 3 & 0 & 0 \end{bmatrix}$$

$$\therefore p = \begin{bmatrix} 0 & -\frac{1}{3} & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

$$x_2 = t_1, x_3 = s_1$$

$$x_1 = -\frac{1}{3}s_1$$

$$x = t_1 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + s_1 \begin{bmatrix} -\frac{1}{3} \\ 0 \\ 1 \end{bmatrix}$$

$$d_1 = 0, x_2 = 0, s_1 = u$$

$$x = u \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\Rightarrow P^{-1} = \begin{bmatrix} 0 & -\frac{1}{3} & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\Rightarrow B = PDP^{-1} = \begin{bmatrix} 0 & -\frac{1}{3} & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ -\frac{1}{3} & 0 & 0 \\ 3 & 0 & 1 \end{bmatrix}$$

Q3

$$P(\text{SPAM}) = 0.2 \quad P(\text{정상}|\text{SPAM}) = 0.5 \quad P(\text{정상}|\text{KSPAM}) = 0.01$$

$$P(\text{K SPAM}) = 0.8 \quad P(\text{정상}|\text{SPAM}) = 0.5 \quad P(\text{정상}|\text{KSPAM}) = 0.99$$

$$P(\text{정상}) = P(\text{정상}|\text{SPAM}) + P(\text{정상}|\text{KSPAM})$$

$$= P(\text{정상}|\text{SPAM}) \times P(\text{SPAM}) + P(\text{정상}|\text{KSPAM}) \times P(\text{KSPAM})$$

$$= 0.5 \times 0.2 + 0.01 \times 0.8$$

베이지안 정리를 이용하면,

$$P(\text{SPAM}|\text{정상}) = \frac{P(\text{정상}|\text{SPAM}) \times P(\text{SPAM})}{P(\text{정상})} = \frac{0.5 \times 0.2}{0.5 \times 0.2 + 0.01 \times 0.8}$$

$$\approx 0.996$$

Q4

$$4-2) \quad H(x) = -\sum_i P(x_i) \log P(x_i)$$

$$= -0.6 \times \log_2 0.6 - 0.4 \times \log_2 0.4$$

$$\approx 0.9170$$

4-3) 4-1)의 야구팀 hit 이산분포 P, 4-3)에 나타낸 야구팀의 hit 이산분포를 구하자면,

$$D_{KL}(P||Q) = \sum_i P(x_i) \log \left(\frac{P(x_i)}{Q(x_i)} \right)$$

$$= 0.6 \times \log_2 \left(\frac{0.6}{0.244} \right) + 0.4 \times \log_2 \left(\frac{0.4}{0.753} \right)$$

$$\approx 0.322$$

Q5)

< Logistic Regression >

 \Rightarrow Loss function : cross entropy

$$\therefore \min_w \int \sum_{i=1}^n (-y_i \log \hat{y}_i - (1-y_i) \log (1-\hat{y}_i)) \quad \hat{y}_i = \frac{1}{1 + e^{-w^T x_i}} = f(w)$$

$$J(w) = -y_i \log \frac{1}{1 + e^{w^T x_i}} - (1-y_i) \log \frac{1}{1 + e^{-w^T x_i}}$$

$$= -y_i \log \frac{1}{1 + e^{w^T x_i}} - (1-y_i) \left(\log \frac{1}{1 + e^{-w^T x_i}} - \log \frac{1}{1 + e^{w^T x_i}} \right)$$

$$= (1-y_i) w^T x_i - (1-y_i) \log \frac{1}{1 + e^{-w^T x_i}} - y_i \log \frac{1}{1 + e^{w^T x_i}}$$

$$= (1-y_i) w^T x_i - \log \frac{1}{1 + e^{-w^T x_i}}$$

since $\nabla^2 f(w) \geq 0$, $J(w)$ is convex

$$\therefore \min_w J(w), w \in \mathbb{R}, J(w) : \text{CONVEX}, \mathbb{R} : \text{CONVEX SET}$$

$$\therefore \nabla J(w) = 0, w = w^* \quad (w^* = \min_w J(w))$$