


MLDL을 위한 수학 과제

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Q1.

$$\bullet A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 3 \\ 0 & 0 & 3 \end{bmatrix}$$

$$(1) \det(A) = \begin{vmatrix} 1 & 0 & 1 \\ 0 & 2 & 3 \\ 0 & 0 & 3 \end{vmatrix} = 1 \begin{vmatrix} 2 & 3 \\ 0 & 3 \end{vmatrix} = 6 \quad \therefore 6$$

기하학적 의미 : 선형 변환을 통해서 basis vector가 바뀌었을 때, 공간이 얼마나 늘어나고 줄어드는지.

basis vector의 직렬으로 이루어진 공간의 넓이 (부피)

(2) Eigenvector : 정방행렬 A 를 선형변환으로 봤을 때, 선형변환 A 에 의한 결과가 자기 자신의 상수 배가 되는 0이 아닌 벡터

$$\Leftrightarrow AV = \lambda V$$

Eigenvalue : λ ; 이 상수배 값을 의미

① Eigenvalue 구하기 : $\det(\lambda I - A) = 0$

$$\det \begin{bmatrix} \lambda - 1 & 0 & -1 \\ 0 & \lambda - 2 & -3 \\ 0 & 0 & \lambda - 3 \end{bmatrix} = (\lambda - 1)(\lambda - 2)(\lambda - 3) = 0$$
$$\therefore \lambda = 1, 2, 3$$

② Eigenvektor 찾기 : $(\lambda I - A)x = 0$

$$\begin{bmatrix} \lambda - 1 & 0 & 1 \\ 0 & \lambda - 2 & 3 \\ 0 & 0 & \lambda - 3 \end{bmatrix}$$

i) $\lambda = 1$

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & -1 & 3 \\ 0 & 0 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_3 \\ -x_2 + 3x_3 \\ -2x_3 \end{bmatrix} \quad \begin{matrix} 3x_3 = x_2 \\ \therefore \end{matrix} \begin{bmatrix} 0 \\ 1 \\ 3 \end{bmatrix}$$

ii) $\lambda = 2$

$$\begin{bmatrix} -1 & 0 & 1 \\ 0 & 0 & 3 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 + x_3 \\ x_3 \\ -x_3 \end{bmatrix} \quad \therefore \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

iii) $\lambda = 3$

$$\begin{bmatrix} -2 & 0 & 1 \\ 0 & -1 & 3 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -2x_1 + x_3 \\ -x_2 + 3x_3 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{matrix} 2x_1 = x_3 \\ x_2 = 3x_3 \end{matrix}$$

② Eigenvektor 찾기 : $(\lambda I - A)x = 0$

$$\begin{bmatrix} \lambda - 1 & 0 & 1 \\ 0 & \lambda - 2 & 3 \\ 0 & 0 & \lambda - 3 \end{bmatrix}$$

i) $\lambda = 1$

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & -1 & 3 \\ 0 & 0 & -2 \end{bmatrix} \sim \begin{bmatrix} 0 & 1 & 3 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{array}{l} x_2 + x_3 = 0 \\ x_3 = 0 \\ x_3 = 0 \end{array} \quad \therefore \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

ii) $\lambda = 2$

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

iii) $\lambda = 3$

$$\begin{bmatrix} 2 & 0 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{bmatrix} \quad \therefore \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix}$$

Q2.

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

• Eigenvalue

$$\begin{vmatrix} \lambda - 0 & 0 & 0 \\ 0 & \lambda - 0 & 0 \\ 3 & 0 & \lambda - 1 \end{vmatrix} = \lambda \begin{vmatrix} \lambda & 0 \\ 0 & \lambda - 1 \end{vmatrix} = \lambda^2(\lambda - 1) = 0 \quad \therefore \lambda = 0, 1$$

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

Q3.

$$P(\text{spam}) = 0.2$$

$$P(\text{당첨} \mid \text{spam}) = 0.5$$

$$P(\text{당첨} \mid \text{not spam}) = 0.01$$

$$* P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

\swarrow spam \swarrow 당첨

$$\Rightarrow P(\text{spam} \mid \text{당첨}) = \frac{(0.5)(0.2)}{0.51} = \frac{0.10}{0.51} = \frac{10}{51} \quad \therefore \frac{10}{51}$$

Q4.

(2) hit : 9 * 이항분포의 entropy : $H(p) = -p \cdot \log_2(p) - (1-p) \cdot \log_2(1-p)$

out : 6

$$p = \frac{9}{15} = \frac{3}{5} = 0.6$$

$$\Rightarrow H(p) = -0.6 \cdot \log_2(0.6) - (1-0.6) \cdot \log_2(0.4)$$

$$1-p = \frac{6}{15} = \frac{2}{5} = 0.4$$

$$\doteq 0.971$$

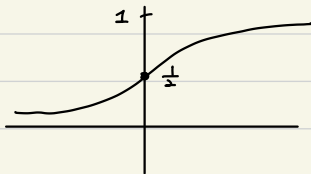
(3) 다른팀 hit = 4
out = 11 } $p = \frac{4}{15} \Rightarrow 1-p = \frac{11}{15}$

* KL Divergence : $KL(p||q) = H(p, q) - H(p) = \sum p_i \log \frac{p_i}{q_i} \quad \because \text{이항분포}$

$$\therefore KL(p||q) = (0.6) \left(\log \frac{0.6}{4/15} \right) \times 15$$

$$= 9 \log \frac{9}{4}$$


Q5.



loss function : $-\log \frac{e^{wTx}}{1+e^{wTx}}$

→ Hessian is positive semi definite

⇒ convex 

∴  global minimum ← GD method 사용.