

## Derivation of Normal Equation

Derive  $\mathbf{w} = (X_b^T X_b)^{-1} X_b^T \mathbf{y}$

เรามี  $Cost = \sum_{i=1}^n (y_i - \hat{y}_i)^2$  และเราต้องการหา  $\mathbf{w} = \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_p \end{bmatrix}$  ที่ทำให้  $Cost$  ต่ำที่สุด

จาก Calculus เราทราบว่า  $Cost$  ต่ำสุดเมื่อ  $\nabla Cost = 0$  ดังนั้น

$$\nabla Cost = \begin{bmatrix} \frac{\partial Cost}{\partial w_0} \\ \frac{\partial Cost}{\partial w_1} \\ \vdots \\ \frac{\partial Cost}{\partial w_d} \\ \vdots \\ \frac{\partial Cost}{\partial w_p} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

พิจารณา  $\frac{\partial Cost}{\partial w_d} = 0$  เมื่อ  $d \in \{0, 1, \dots, p\}$  จะได้ว่า

$$\sum_{i=1}^n x_{i,d} \hat{y}_i = \sum_{i=1}^n x_{i,d} y_i \quad (1)$$

ซึ่งสามารถเขียนให้อยู่ในรูป Matrix ได้ดังนี้

$$\begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} = \begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \quad (2)$$

จากสมการ

$$\begin{bmatrix} \frac{\partial Cost}{\partial w_0} \\ \frac{\partial Cost}{\partial w_1} \\ \vdots \\ \frac{\partial Cost}{\partial w_d} \\ \vdots \\ \frac{\partial Cost}{\partial w_p} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

เราจะสามารถเขียนให้อยู่ในรูปต่อไปนี้ได้

$$\begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{1,1} & x_{2,1} & \cdots & x_{n,1} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1,p} & x_{2,p} & \cdots & x_{n,p} \end{bmatrix} \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{1,1} & x_{2,1} & \cdots & x_{n,1} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1,p} & x_{2,p} & \cdots & x_{n,p} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} ; x_{i,0} = 1; i \in \{1, 2, \dots, n\}$$

กำหนดให้

$$X_b = \begin{bmatrix} 1 & x_{1,1} & \cdots & x_{1,p} \\ 1 & x_{2,1} & \cdots & x_{2,p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n,1} & \cdots & x_{n,p} \end{bmatrix}, \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \hat{\mathbf{y}} = \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix}$$

นั่นคือ  $X_b^T \hat{\mathbf{y}} = X_b^T \mathbf{y}$

$$\therefore \hat{\mathbf{y}} = X_b \mathbf{w} \tag{3}$$

จะได้  $X_b^T X_b \mathbf{w} = X_b^T \mathbf{y}$

ดังนั้น  $\mathbf{w} = (X_b^T X_b)^{-1} X_b^T \mathbf{y}$

Derive Equation 1

$$\begin{aligned}
\frac{\partial Cost}{\partial w_d} &= \frac{\partial \sum_{i=1}^n (y_i - \hat{y}_i)^2}{\partial w_d} \\
0 &= \sum_{i=1}^n \frac{\partial (y_i - \hat{y}_i)^2}{\partial (y_i - \hat{y}_i)} \frac{\partial (y_i - \hat{y}_i)}{\partial \hat{y}_i} \frac{\partial \hat{y}_i}{\partial w_d} \\
0 &= \sum_{i=1}^n 2(y_i - \hat{y}_i)(1)(-1) \frac{\partial \hat{y}_i}{\partial w_d} \\
0 &= \sum_{i=1}^n -2(y_i - \hat{y}_i) \frac{\partial (w_0 + w_1 x_{i,1} + \dots + w_d x_{i,d} + \dots + w_p x_{i,p})}{\partial w_d} \\
0 &= \sum_{i=1}^n -2(y_i - \hat{y}_i) x_{i,d} \\
0 &= -2 \sum_{i=1}^n (y_i - \hat{y}_i) x_{i,d} \\
0 &= \sum_{i=1}^n (y_i - \hat{y}_i) x_{i,d} \\
0 &= \sum_{i=1}^n x_{i,d} (y_i - \hat{y}_i) \\
0 &= \sum_{i=1}^n x_{i,d} y_i - \sum_{i=1}^n x_{i,d} \hat{y}_i \\
0 &= \sum_{i=1}^n x_{i,d} y_i - \sum_{i=1}^n x_{i,d} \hat{y}_i \\
\sum_{i=1}^n x_{i,d} \hat{y}_i &= \sum_{i=1}^n x_{i,d} y_i
\end{aligned}$$

Derive Equation 2

$$\text{จาก } \sum_{i=1}^n x_{i,d} \hat{y}_i = \sum_{i=1}^n x_{i,d} y_i$$

$$\text{พิจารณา } \sum_{i=1}^n x_{i,d} \hat{y}_i$$

$$\sum_{i=1}^n x_{i,d} \hat{y}_i = x_{1,d} \hat{y}_1 + x_{2,d} \hat{y}_2 + \cdots + x_{n,d} \hat{y}_n$$

$$= \begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix}$$

$$\text{พิจารณา } \sum_{i=1}^n x_{i,d} y_i$$

$$\sum_{i=1}^n x_{i,d} y_i = x_{1,d} y_1 + x_{2,d} y_2 + \cdots + x_{n,d} y_n$$

$$= \begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

Derive Equation 3

$$\begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} = \begin{bmatrix} w_0 + w_1x_{1,1} + \cdots + w_px_{1,p} \\ w_0 + w_1x_{2,1} + \cdots + w_px_{2,p} \\ \vdots \\ w_0 + w_1x_{n,1} + \cdots + w_px_{n,p} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & x_{1,1} & \cdots & x_{1,p} \\ 1 & x_{2,1} & \cdots & x_{2,p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n,1} & \cdots & x_{n,p} \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_p \end{bmatrix}$$

กำหนดให้  $\hat{\mathbf{y}} = \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix}$ ,  $X_b = \begin{bmatrix} 1 & x_{1,1} & \cdots & x_{1,p} \\ 1 & x_{2,1} & \cdots & x_{2,p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n,1} & \cdots & x_{n,p} \end{bmatrix}$ ,  $\mathbf{w} = \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_p \end{bmatrix}$

จะได้ว่า  $\hat{\mathbf{y}} = X_b \mathbf{w}$