## Derivation of Elastic Net

Derive 
$$\mathbf{w} = (X_b^T X_b + \lambda I)^{-1} X_b^T \mathbf{y}$$
 เรามี  $Cost = \frac{1}{2} \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda (l1_{ratio}) \sum_{j=0}^p |w_j| + \frac{1}{2} \lambda (1 - l1_{ratio}) \sum_{j=0}^p w_j^2$  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  $\dfrac{\partial Cost}{\partial w_d}=0$  เมื่อ  $d\in\{0,1,...,p\}$  จะได้ว่า

$$\sum_{i=1}^{n} x_{i,d} \hat{y}_i + \lambda (l1_{ratio}) \frac{\partial |w_d|}{\partial w_d} + \lambda (1 - l1_{ratio}) w_d = \sum_{i=1}^{n} x_{i,d} y_i$$
 (1)

ซึ่งสามารถเขียนให้อยู่ในรูป matrix ได้ดังนี้

$$\begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} + \lambda (l1_{ratio}) \frac{\partial |w_d|}{\partial w_d} + \lambda (1 - l1_{ratio}) w_d = \begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$
(2)

จากสมการ

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

เราจะสามารถเขียนให้อยู่ในรูปต่อไปนี้ได้

$$\begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{1,1} & x_{2,1} & \cdots & x_{n,1} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1,p} & x_{2,p} & \cdots & x_{n,p} \end{bmatrix} \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} + \lambda(l1_{ratio}) \begin{bmatrix} \frac{\partial |w_0|}{\partial w_0} \\ \frac{\partial |w_1|}{\partial w_1} \\ \vdots \\ \frac{\partial |w_p|}{\partial w_n} \end{bmatrix} + \lambda(1-l1_{ratio}) \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_p \end{bmatrix} = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{1,1} & x_{2,1} & \cdots & x_{n,1} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1,p} & x_{2,p} & \cdots & x_{n,p} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

โดยที่  $x_{i,0}=1; i\in\{1,2,...,n\}$ 

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  $X_h^T \hat{\mathbf{y}} + \lambda \mathbf{w} = X_h^T \mathbf{y}$ 

จะได้

$$X_b^T X_b \mathbf{w} + \lambda (l1_{ratio}) \nabla \mathbf{w} + \lambda (1 - l1_{ratio}) \mathbf{w} = X_b^T \mathbf{y}$$

## Derive Equation 1

เนื่องจาก

$$\frac{\partial \frac{1}{2} \sum_{i=1}^{n} (y_{i} - \hat{y}_{i})^{2}}{\partial w_{d}} = \frac{1}{2} \sum_{i=1}^{n} \frac{\partial (y_{i} - \hat{y}_{i})^{2}}{\partial (y_{i} - \hat{y}_{i})} \cdot \frac{\partial (y_{i} - \hat{y}_{i})}{\partial \hat{y}_{i}} \cdot \frac{\partial \hat{y}_{i}}{w_{d}}$$

$$= \frac{1}{2} \sum_{i=1}^{n} \frac{\partial (y_{i} - \hat{y}_{i})^{2}}{\partial (y_{i} - \hat{y}_{i})} \cdot \frac{\partial (y_{i} - \hat{y}_{i})}{\partial \hat{y}_{i}} \cdot \frac{\partial \hat{y}_{i}}{w_{d}}$$

$$= \frac{1}{2} \sum_{i=1}^{n} 2(y_{i} - \hat{y}_{i})(1)(-1)\frac{\partial \hat{y}_{i}}{w_{d}}$$

$$= \frac{1}{2} \sum_{i=1}^{n} -2(y_{i} - \hat{y}_{i})\frac{\partial (w_{0} + w_{1}x_{i,1} + \dots + w_{d}x_{i,d} + \dots + w_{p}x_{i,p})}{w_{d}}$$

$$= \frac{1}{2} \sum_{i=1}^{n} -2(y_{i} - \hat{y}_{i})x_{i,d}$$

$$= \frac{1}{2} \sum_{i=1}^{n} (y_{i} - \hat{y}_{i})x_{i,d}$$

$$= -\sum_{i=1}^{n} (y_{i} - \hat{y}_{i})x_{i,d}$$

$$= -\sum_{i=1}^{n} x_{i,d}y_{i} + \sum_{i=1}^{n} x_{i,d}\hat{y}_{i}$$

และ

$$\begin{split} \frac{\partial \lambda(l1_{ratio}) \sum_{j=0}^{p} |w_{j}|}{\partial w_{d}} &= \lambda(l1_{ratio}) \frac{\partial \sum_{j=0}^{p} |w_{j}|}{\partial w_{d}} \\ &= \lambda(l1_{ratio}) \frac{\partial (|w_{0}| + |w_{1}| + \dots + |w_{d}| + \dots + |w_{p}|)}{\partial w_{d}} \\ &= \lambda(l1_{ratio}) \frac{\partial |w_{d}|}{\partial w_{d}} \end{split}$$

และ

$$\begin{split} \frac{\partial \frac{1}{2}\lambda(1-l1_{ratio})\sum_{j=0}^{p}w_{j}^{2}}{\partial w_{d}} &= \frac{1}{2}\lambda(1-l1_{ratio})\frac{\partial\sum_{j=0}^{p}w_{j}^{2}}{\partial w_{d}} \\ &= \frac{1}{2}\lambda(1-l1_{ratio})\frac{\partial(w_{0}^{2}+w_{1}^{2}+\cdots+w_{d}^{2}+\cdots+w_{p}^{2})}{\partial w_{d}} \\ &= \frac{1}{2}\lambda(1-l1_{ratio})2w_{d} \\ &= \frac{2}{2}\lambda(1-l1_{ratio})w_{d} \\ &= \lambda(1-l1_{ratio})w_{d} \end{split}$$

ดังนั้น

$$\begin{split} \frac{\partial Cost}{\partial w_d} &= \frac{\partial \left[\frac{1}{2}\sum_{i=1}^n \left(y_i - \hat{y}_i\right)^2 + \lambda(l1_{ratio})\sum_{j=0}^p |w_j| + \frac{1}{2}\lambda(1 - l1_{ratio})\sum_{j=0}^p w_j^2\right]}{\partial w_d} \\ 0 &= \frac{\partial \frac{1}{2}\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\partial w_d} + \frac{\partial \lambda(l1_{ratio})\sum_{j=0}^p |w_j|}{\partial w_d} + \frac{\partial \frac{1}{2}\lambda(1 - l1_{ratio})\sum_{j=0}^p w_j^2}{\partial w_d} \\ 0 &= -\sum_{i=1}^n x_{i,d}y_i + \sum_{i=1}^n x_{i,d}\hat{y}_i + \lambda(l1_{ratio})\frac{\partial |w_d|}{\partial w_d} + \lambda(1 - l1_{ratio})w_d \end{split}$$

นั่นคือ

$$\sum_{i=1}^{n} x_{i,d} \hat{y}_i + \lambda(l1_{ratio}) \frac{\partial |w_d|}{\partial w_d} + \lambda(1 - l1_{ratio}) w_d = \sum_{i=1}^{n} x_{i,d} y_i$$

Derive Equation 2

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

$$\sum_{i=1}^{n} x_{i,d} \hat{y}_{i} = x_{1,d} \hat{y}_{1} + x_{2,d} \hat{y}_{2} + \dots + x_{n,d} \hat{y}_{n}$$

$$= \begin{bmatrix} x_{1,d} & x_{2,d} & \dots & x_{n,d} \end{bmatrix} \begin{bmatrix} \hat{y}_{1} \\ \hat{y}_{2} \\ \vdots \\ \hat{y}_{n} \end{bmatrix}$$

พิจารณา  $\sum_{i=1}^n x_{i,d} y_i$ 

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

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

ดังนั้น จะได้ว่า

$$\sum_{i=1}^{n} x_{i,d} \hat{y}_i + \lambda (l1_{ratio}) \frac{\partial |w_d|}{\partial w_d} + \lambda (1 - l1_{ratio}) w_d = \sum_{i=1}^{n} x_{i,d} y_i$$

$$\begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} + \lambda (l1_{ratio}) \frac{\partial |w_d|}{\partial w_d} + \lambda (1 - l1_{ratio}) w_d = \begin{bmatrix} x_{1,d} & x_{2,d} & \cdots & x_{n,d} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

## Derive Equation 3

$$\begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \vdots \\ \hat{y}_n \end{bmatrix} = \begin{bmatrix} w_0 + w_1 x_{1,1} + \dots + w_p x_{1,p} \\ w_0 + w_1 x_{2,1} + \dots + w_p x_{2,p} \\ \vdots \\ w_0 + w_1 x_{n,1} + \dots + w_p x_{n,p} \end{bmatrix}$$
$$= \begin{bmatrix} 1 & x_{1,1} & \dots & x_{1,p} \\ 1 & x_{2,1} & \dots & x_{2,p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n,1} & \dots & x_{n,p} \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_p \end{bmatrix}$$

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

ดังนั้น จะได้ว่า

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