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1. public private

(1) 5.54 5.63

(2) 6.05 5.98

若把預測值設為前一小時的PM2.5, RMSE為6.4

用前9H的PM2.5做feature為6.05提升不大, 由此可知時間

遠近影響很大, 加其它 feature的提升也不是很顯著

2. 去除PM2.5值 &lt; 2 或 &gt; 100 的 data

去除風向和小時平均風向這2個變數: RMSE從7.47 → 4.70

使用前4小時來預測就好: RMSE = 3.9

用5小時以上RMSE明顯上升

testing data 若有漏缺, 補上該變數的平均值

1-(a)

$$L_{ssq}(w, b) = \frac{1}{10} [(1.2 - w - b)^2 + (2.4 - 2w - b)^2 + (3.5 - 3w - b)^2 + (4.1 - 4w - b)^2 + (5.6 - 5w - b)^2]$$

$$\text{求 } \frac{\partial L_{ssq}}{\partial w} = 0$$

$$\Rightarrow \frac{1}{5} [(w + b - 1.2) + (4w + 2b - 4.8) + (9w + 3b - 10.5) + (16w + 4b - 16.6) + (25w + 5b - 28)]$$

$$\Rightarrow \frac{1}{5} (55w + 15b - 61.1) = 0$$

$$\text{求 } \frac{\partial L_{ssq}}{\partial b} = 0$$

$$\Rightarrow \frac{1}{5} (5b + 15w - 16.8) = 0$$

$$\begin{cases} 55w + 15b - 61.1 = 0 \\ 45w + 15b - 50.4 = 0 \end{cases}$$

$$10w = 10.7$$

$$w = 1.07$$

$$b = 0.15$$

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$$1-(b) L_{sq}(w, b) = \frac{1}{2N} \sum_{i=1}^N (y_i - (w^T x_i + b))^2$$

$$\frac{\partial L}{\partial b} = 0$$

$$\Rightarrow \frac{1}{N} \sum_{i=1}^N (y_i - w^T x_i - b) = 0 \Rightarrow \sum y_i - \sum w^T x_i - \sum b = 0$$

$$\Rightarrow N \cdot b = \sum y_i - w^T \sum x_i \Rightarrow b = \bar{y} - w^T \bar{x}$$

$$\frac{\partial L}{\partial w} = 0$$

$$\Rightarrow \frac{1}{N} \sum_{i=1}^N (y_i - w^T x_i - b) x_i = 0 \Rightarrow \sum y_i x_i - w \sum x_i^2 - \boxed{b \sum x_i} = 0$$

$$\Rightarrow \sum y_i x_i - w \sum x_i^2 - (\bar{y} - w^T \bar{x}) \sum x_i = 0$$

$$\Rightarrow w \sum (x_i^T \bar{x} - x_i^2) = \bar{y} \sum x_i - \sum y_i x_i = \sum x_i (\bar{y} - \bar{y})$$

$$\Rightarrow w = \frac{\sum_{i=1}^N x_i (\bar{y} - y_i)}{\sum_{i=1}^N x_i (\bar{x} - x_i)}$$

$$1-(c) \quad L_{\text{reg}}(w, b) = \frac{1}{2N} \sum_{i=1}^N (y_i - (w^T x_i + b))^2 + \frac{\lambda}{2} \|w\|^2, \lambda \geq 0$$

$$\frac{\partial L_{\text{reg}}}{\partial b} = 0 \Rightarrow \text{同 } 1-(b): \quad b = \bar{y} - w^T \bar{x}$$

$$\frac{\partial L_{\text{reg}}}{\partial w} = 0$$

$$\Rightarrow \sum y_i x_i - w \sum x_i^2 - (\bar{y} - w^T \bar{x}) \sum x_i + \lambda w = 0$$

$$\Rightarrow w (\lambda + \sum (x_i^T \bar{x} - x_i^2)) = \sum x_i (\bar{y} - y_i)$$

$$\Rightarrow w = \frac{\sum_{i=1}^N x_i (\bar{y} - y_i)}{\lambda + \sum_{i=1}^N x_i^T (\bar{x} - x_i)}$$

$$2. \quad L_{\text{noise}}(w, b) = \frac{1}{2N} \sum_{i=1}^N (y_i - w^T (x_i + \eta_i) - b)^2, \sigma^2 = E[\eta_i^2]$$

$$\frac{\partial L_{\text{noise}}}{\partial b} = 0 \Rightarrow \frac{1}{N} \sum_{i=1}^N (y_i - w^T x_i - w^T \eta_i - b) = 0$$

$$\Rightarrow \sum y_i - w^T \sum x_i - w^T \underbrace{\sum \eta_i}_{=0} - N \cdot b = 0 \Rightarrow b = \bar{y} - w^T \bar{x}$$

$$\frac{\partial L_{\text{noise}}}{\partial w} = 0 \Rightarrow w = \frac{\sum (x_i + \eta_i) (y_i - \bar{y})}{\sum (x_i + \eta_i)^T (x_i + \eta_i - \bar{x})}$$

$$w = \frac{\sum x_i (y_i - \bar{y}) - \bar{y} \sum \eta_i + \sum \eta_i y_i}{\sum x_i^T (x_i - \bar{x}) + x_i^T \eta_i + \eta_i^T x_i + \eta_i^T \eta_i - \bar{x} \eta_i^T}$$

$$w = \frac{\sum x_i (y_i - \bar{y}) - 0 + 0}{\sum x_i^T (x_i - \bar{x}) + 0 + 0 + \sigma^2 - 0}, \text{ 等價於 } 1-(c) \quad \lambda = \sigma^2$$

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$$3-(a) \sum_{i=1}^N g_k(x_i) y_i = A$$

$$\begin{cases} e_0 = \frac{1}{N} \sum y_i^2 \\ e_k = \frac{1}{N} \sum (g_k(x_i) - y_i)^2 = \frac{1}{N} \sum (g_k(x_i)^2 - 2y_i g_k(x_i) + y_i^2) \\ s_k = \frac{1}{N} \sum (g_k(x_i))^2 \end{cases}$$

$$e_k = s_k - 2A + e_0 \Rightarrow A = \frac{s_k + e_0 - e_k}{2} \quad \#$$

$$3-(b) L = \min \left[ \frac{1}{N} \sum_{i=1}^N \left( \sum_{k=1}^K \alpha_k g_k(x_i) - y_i \right)^2 \right]$$

$$\frac{\partial L}{\partial \alpha_k} = \frac{2}{N} \sum_{i=1}^N (\alpha_k g_k(x_i) - y_i) g_k(x_i) = 0$$

$$\Rightarrow \alpha_k g_k(x_i) - y_i = 0 \quad \checkmark \quad g_k(x_i) = 0$$

$$\Rightarrow \alpha_k = \frac{y_i}{g_k(x_i)} \quad \#$$