Data. ML. 100, Homework, Week 1

$$\nabla \mathcal{E}_{MSE} = \overline{0} \implies \left(\frac{\partial \mathcal{E}_{MSE}}{\partial a}\right) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \begin{array}{c} \text{Chain rule: } f \rightarrow z^2 \\ \frac{\partial \mathcal{E}_{MSE}}{\partial b} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \begin{array}{c} \text{Spanson} \\ \text{f(g(x))}, & \text{Spanson} \\ \text{fig(x)} \end{pmatrix}, \quad \begin{array}{c} \text{Spanson} \\ \text{Spanson} \\ \text{Spanson} \end{array}$$

$$\frac{\partial \mathcal{E}_{MSE}}{\partial \alpha} = \frac{\partial \frac{1}{N} \sum (y_i - \hat{y}_i)^2}{\partial \alpha} \Rightarrow \frac{\partial \frac{1}{N} \sum (y_i - \alpha x_i - b)^2}{\partial \alpha} = 0$$

$$\Rightarrow \frac{1}{N} \cdot 2 \cdot \sum (y_i - \alpha x_i - b) \cdot (-x_i) = 0 \iff \alpha = \frac{\sum y_i x_i - b \sum x_i}{\sum_{i=1}^{N} y_i - k}$$

 $\Rightarrow \frac{1}{N} \cdot 2 \cdot \sum (y_i - ax_i - b) \cdot (-x_i) = 0 \iff \alpha = \frac{\sum y_i x_i - b \sum x_i}{\sum x_i^2}$   $\text{Subst. 1° to } \frac{\partial \mathcal{E}_{MSE}}{\partial b} \Rightarrow \text{Solve } b \Rightarrow \text{back Subst } b \text{ to 1°} \Rightarrow \text{Solve } \alpha$ 

$$\frac{\partial \mathcal{E}_{MSE}}{\partial b} \Rightarrow \frac{\partial \frac{1}{N} \Sigma (y_i - \alpha x_i - b)^2}{\partial b} = 0 \Rightarrow \frac{\partial \frac{1}{N} \Sigma (y_i - \frac{\Sigma y_i x_i - b \Sigma x_i}{\Sigma x_i^2} x_i - b)^2}{\partial b} = 0$$

$$\Rightarrow \frac{1}{N} \cdot \left( -\frac{\sum x_i}{\sum x_i^2} - 1 \right) \cdot 2 \sum \left( y_i - \frac{\sum y_i x_i - b \sum x_i}{\sum x_i^2} x_i - b \right) = 0 \quad 1 : \frac{1}{N} \cdot \left( -\frac{\sum x_i}{\sum x_i^2} - 1 \right) \cdot 2$$

$$= L(N\Sigma x_i^2 - (\Sigma x_i)^2) = \Sigma y_i \Sigma x_i^2 - \Sigma y_i x_i \Sigma x_i + [\Sigma x_i^2 (N - (\Sigma x_i)^2)]$$

$$b = \frac{\sum y_i \sum x_i^2 - \sum y_i x_i \sum x_i}{N \sum x_i^2 - (\sum x_i)^2}$$
 subst b back to 1° to solve a

$$\alpha = \frac{\sum y_i x_i}{\sum x_i^2} - \frac{\sum y_i \sum x_i^2 - \sum y_i x_i \sum x_i}{N \sum x_i^2 - (\sum x_i)^2} \cdot \frac{\sum x_i}{\sum x_i^2}$$

$$\alpha = \frac{\sum y_i x_i}{\sum x_i^2} - \frac{\sum y_i \sum x_i^2 \sum x_i - \sum y_i x_i \sum x_i}{\sum x_i^2 \left(N \sum x_i^2 - \left(\sum x_i\right)^2\right)}$$

$$\alpha = \frac{\sum y_i \times i}{\sum x_i^2} - \frac{\sum y_i \sum x_i^2 \left(N \sum x_i - \sum y_i \times i \left(\sum x_i\right)^2}{\sum x_i^2 \left(N \sum x_i^2 - \left(\sum x_i\right)^2\right)}$$