

1. a)  $y = \theta_0 + \theta_1 x$ .

$$X = \begin{bmatrix} 1 \\ 3 \\ 6 \end{bmatrix} \quad Y = \begin{bmatrix} 6 \\ 10 \\ 16 \end{bmatrix}.$$

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$$\hat{y}_i = \theta_0 + \theta_1 x_i$$

using  
Normal Eq.  
Method.

$$\begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \hat{y}_3 \end{bmatrix} = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \end{bmatrix} \begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix}$$

$$\hat{Y} = X \theta$$

$$\text{error } (e_i) = y_i - \hat{y}_i$$

$$\text{let } e = \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}$$

$$Y = X\theta + e \quad \text{--- (1)}$$

we need to minimize  $e_1^2 + e_2^2 + e_3^2$

$\Rightarrow$  we need to minimize  $e^T e$ .

$$e = Y - X\theta \quad (\text{from (1)}).$$

$$\varepsilon_1^T \varepsilon_1 = (Y - X\theta)^T (Y - X\theta).$$

$$\varepsilon_1^T \varepsilon_1 = (Y^T - \theta^T X^T) (Y - X\theta)$$

$$\varepsilon_1^T \varepsilon_1 = Y^T Y - Y^T X \theta - \theta^T X^T Y + \theta^T X^T X \theta.$$

$$\varepsilon_1^T \varepsilon_1 = Y^T Y - 2Y^T X \theta + \theta^T X^T X \theta$$

$$\frac{\partial}{\partial \theta} (\varepsilon_1^T \varepsilon_1) = 0 - 2X^T Y + 2X^T X \theta = 0$$

$$2X^T Y = 2X^T X \theta$$

$$X^T Y = X^T X \theta$$

$$\theta = (X^T X)^{-1} X^T Y.$$

$$X^T X = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 3 & 6 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 3 \\ 1 & 6 \end{bmatrix}$$

$$X^T X = \begin{bmatrix} 3 & 10 \\ 10 & 46 \end{bmatrix}$$

$$\begin{aligned} |X^T X| &= 3 \times 46 - 100 = 138 - 100 \\ &= 38, \end{aligned}$$

$$\text{adj}(X^T X) = \begin{bmatrix} 46 & -10 \\ -10 & 3 \end{bmatrix}$$

$$(X^T X)^{-1} = \frac{\text{adj}(X^T X)}{|X^T X|} = \frac{1}{38} \begin{bmatrix} 46 & -10 \\ -10 & 3 \end{bmatrix}.$$

$$\Theta = (X^T X)^{-1} X^T Y.$$

$$\Theta = \frac{1}{38} \begin{bmatrix} 46 & -10 \\ -10 & 3 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 3 & 6 \end{bmatrix}_{2 \times 3} \begin{bmatrix} 6 \\ 10 \\ 16 \end{bmatrix}_{3 \times 1}$$

$$\Theta = \frac{1}{38} \begin{bmatrix} 46 & -10 \\ -10 & 3 \end{bmatrix} \begin{bmatrix} 32 \\ 132 \end{bmatrix}$$

$$= \frac{1}{38} \begin{bmatrix} 46 \times 32 - 1320 \\ -320 + 132 \times 3 \end{bmatrix}$$

$$= \frac{1}{38} \begin{bmatrix} 152 \\ 76 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix} = \begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix}.$$

$$b) \quad X = \begin{bmatrix} 1 \\ 3 \\ 6 \end{bmatrix} \quad Y = \begin{bmatrix} 6 \\ 10 \\ 16 \end{bmatrix}$$

$$y_i^o = \theta_0 + \theta_1 x_i^o.$$

initial Guess is  $(\theta_0, \theta_1) = (0, 0)$ .

$$\hat{Y} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$e_i^o = y_i - \hat{y}_i$$

$$MSE = \frac{\sum_{i=1}^N |e_i^o|^2}{N} \Rightarrow MSE = \frac{|e_1|^2 + |e_2|^2 + |e_3|^2}{3}$$

$$= \frac{(y_1 - (\theta_0 + \theta_1 x_1))^2 + (y_2 - (\theta_0 + \theta_1 x_2))^2 + (y_3 - (\theta_0 + \theta_1 x_3))^2}{3}.$$

$$= \frac{1}{3} \left[ \sum_{i=1}^3 (y_i^2 + (\theta_0 + \theta_1 x_i^o)^2 - 2y_i^o(\theta_0 + \theta_1 x_i^o)) \right]$$

$$\frac{\partial MSE}{\partial \theta_0} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^o) - 2y_i^o) \right]$$

$$\frac{\partial MSE}{\partial \theta_1} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^o)(x_i^o) - 2y_i^o x_i^o) \right].$$

Initial guess  $(\theta_0, \theta_1) = (0, 0)$  First Iteration

$$\Rightarrow \frac{\partial \text{MSE}}{\partial \theta_0} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0) - 2y_i^0) \right]$$

$$= \frac{1}{3} \left[ \sum_{i=1}^3 (2(0 + 0x_i^0) - 2y_i^0) \right]$$

$$= \frac{1}{3} \left[ \sum_{i=1}^3 (-2y_i^0) \right]$$

$$= -\frac{2}{3} \left[ \sum_{i=1}^3 y_i^0 \right] = -\frac{2}{3} [6 + 10 + 16]$$

$$= -\frac{2}{3} (32)$$

$$= -\frac{64}{3}$$

$$\frac{\partial \text{MSE}}{\partial \theta_1} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0)(x_i^0) - 2y_i^0 x_i^0) \right]$$

$$= \frac{1}{3} \left[ \sum_{i=1}^3 (2(0)(x_i^0) - 2y_i^0 x_i^0) \right]$$

$$= -\frac{2}{3} \left[ \sum_{i=1}^3 x_i^0 y_i^0 \right]$$

$$= -\frac{2}{3} [(1)(6) + (3)(10) + (6)(16)]$$

$$= -\frac{2}{3} [6 + 30 + 96] = \frac{132}{3} (-2) = -88$$

updating  $\theta_0, \theta_1$

$$\theta_0 = \theta_0 - \alpha \frac{\partial (\text{MSE})}{\partial \theta_0}$$

$$= 0 - (0.1) \left[ -\frac{64}{3} \right] = 2.134$$

$$\theta_1 = \theta_1 - \alpha \frac{\partial (\text{MSE})}{\partial \theta_1}$$

$$= 0 - (0.1) (-88) = 8.8.$$

$$(\theta_0, \theta_1) = (2.134, 8.8).$$

Second Iteration

$$\frac{\partial (\text{MSE})}{\partial \theta_0} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0) - 2y_i^0) \right]$$

$$= 41.6$$

$$\frac{\partial (\text{MSE})}{\partial \theta_1} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0)(x_i^0) - 2y_i^0 x_i^0) \right]$$

$$= \frac{1}{3} \left[ 2 \left[ 2.134 \times 10 + 8.8 [46] \right] - 2(132) \right]$$

$$= \frac{2}{3} [21.34 + 404.8 - 132] = 196.09.$$

updating  $(\theta_0, \theta_1)$ .

$$\begin{aligned}\theta_0 &= \theta_0 - \alpha \left( \frac{\partial \text{MSE}}{\partial \theta_0} \right) \\ &= 2.134 - (0.1)(41.6) \\ &= -2.02667\end{aligned}$$

$$\begin{aligned}\theta_1 &= \theta_1 - \alpha \left( \frac{\partial \text{MSE}}{\partial \theta_1} \right) \\ &= 8.8 - (0.1)(196.09) \\ &= 8.8 - 19.609 \\ &= -10.809.\end{aligned}$$

$$(\theta_0, \theta_1) = (-2.026, -10.809) \quad \text{Third iteration}$$

$$\begin{aligned}\frac{\partial \text{MSE}}{\partial \theta_0} &= \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0) - 2y_i^0) \right] \\ &= -97.44\end{aligned}$$

$$\begin{aligned}\frac{\partial \text{MSE}}{\partial \theta_1} &= \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0)(x_i^0) - 2y_i^0 x_i^0) \right] \\ &= -432.984\end{aligned}$$

$$\begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix} = \begin{bmatrix} -2.0267 \\ -10.8089 \end{bmatrix} - 0.01 \begin{bmatrix} -97.44 \\ -432.984 \end{bmatrix}$$

$$= \begin{bmatrix} 7.7 \\ 32.4895 \end{bmatrix}$$

$$(\theta_0, \theta_1) = (7.7, 32.4895)$$

Fourth iteration

$$\frac{\partial (MSE)}{\partial \theta_0} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0) - 2y_i^0) \right]$$

$$= 210.7$$

$$\frac{\partial (MSE)}{\partial \theta_1} = \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0)(x_i^0) - 2y_i^0 x_i^0) \right]$$

$$= 959.797$$

$$\begin{pmatrix} \theta_0 \\ \theta_1 \end{pmatrix} = \begin{pmatrix} 7.7 \\ 32.4895 \end{pmatrix} - 0.01 \begin{pmatrix} 210.7 \\ 959.797 \end{pmatrix}$$



$$\begin{pmatrix} \theta_0 \\ \theta_1 \end{pmatrix} = \begin{pmatrix} -13.35 \\ -63.49 \end{pmatrix}$$

$$(\theta_0, \theta_1) = (-13.35, -63.49) \quad \text{fifth iteration.}$$

$$\begin{aligned} \frac{\partial (MSE)}{\partial \theta_0} &= \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0) - 2y_i^0) \right] \\ &= -471.3054. \end{aligned}$$

$$\begin{aligned} \frac{\partial (MSE)}{\partial \theta_1} &= \frac{1}{3} \left[ \sum_{i=1}^3 (2(\theta_0 + \theta_1 x_i^0)(x_i^0) - 2y_i^0 x_i^0) \right] \\ &= -2124.0464. \end{aligned}$$

$$\underline{\underline{\begin{pmatrix} \theta_0 \\ \theta_1 \end{pmatrix} = \begin{pmatrix} 33.7786 \\ 148.914 \end{pmatrix}}}$$

$$(c) \quad \theta_1 = \frac{\text{Cov}(X, Y)}{\text{Var}(X)}.$$

$$\theta_0 = \bar{y} - \theta_1 \bar{x}$$

$$\text{Cov}(X, Y) = E(XY) - E[X]E[Y].$$

$$E[X] = \frac{1+3+6}{3} = 10/3.$$

$$E[Y] = \frac{6+10+16}{3} = \frac{32}{3}$$

$$E[XY] = \frac{132}{3}.$$

$$\begin{aligned} \text{Cov}(X, Y) &= E(XY) - E(X)E(Y) \\ &= \frac{132}{3} - \frac{10}{3} \times \frac{32}{3} \\ &= \frac{76}{9}. \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &= E(X^2) - (E(X))^2 \\ &= \frac{1+9+36}{3} - \left(\frac{10}{3}\right)^2 \\ &= 38/9 \end{aligned}$$

$$\theta_1 = \frac{\text{Cov}(X, Y)}{\text{Var}(X)} = \underline{\underline{2}}$$

$$\theta_0 = E[Y] - \theta_1 E[X]$$

$$= \frac{32}{3} - \frac{2 \times 10}{3}$$

$$= \frac{12}{3} = 4$$

$$(\theta_0, \theta_1) = \underline{\underline{(4, 2)}}$$

$$2. (b) \quad X = \begin{bmatrix} 1 & 2 \\ 2 & 4 \\ 3 & 6 \\ 4 & 8 \end{bmatrix}$$

$$Y = \begin{bmatrix} 2 \\ 3 \\ 4 \\ 5 \end{bmatrix}$$

$$(X^T X)^{-1} X^T Y = \theta.$$

$$\text{where } X = \begin{bmatrix} 1 & 1 & 2 \\ 1 & 2 & 4 \\ 1 & 3 & 6 \\ 1 & 4 & 8 \end{bmatrix}$$

$$X^T X = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 2 & 4 & 6 & 8 \end{bmatrix} \begin{bmatrix} 1 & 1 & 2 \\ 2 & 4 & 6 \\ 3 & 6 & 8 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 10 & 20 \\ 10 & 30 & 60 \\ 20 & \underline{60} & \underline{120} \end{bmatrix}$$

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↳ dependent on each other.

⇒ Multicollinearity ⇒  $X^T X$  is not invertible