

① a) we can solve that using normal equations:

$$A = X^T X = \begin{pmatrix} 1.5 & 3 & 4.5 \\ 2 & 2.5 & 3 \end{pmatrix} \begin{pmatrix} 1.5 & 2 \\ 3 & 2.5 \\ 4.5 & 3 \end{pmatrix} = \begin{pmatrix} 31.5 & 24 \\ 24 & 17.25 \end{pmatrix}$$

$$B = X^T Y = \begin{pmatrix} 1.5 & 3 & 4.5 \\ 2 & 2.5 & 3 \end{pmatrix} \begin{pmatrix} 10 \\ 15.5 \\ 21 \end{pmatrix} = \begin{pmatrix} 156 \\ 121.75 \end{pmatrix}$$

$$\text{Solve } Ax = B \rightarrow x = A^{-1} B = \begin{pmatrix} 2.6 \\ 3 \end{pmatrix} = \beta$$

② First of all, in the Part a), the analytical solution seems easy to find. But it is not the case for all ML problems. Some cases the analytical solution might be very complex to find. The second reason can be that in such cases, these kinds of solutions might be computationally infeasible for current available compute powers. But learning algorithms can be implemented without this issue. Moreover, they can also be designed to be flexible.

$$\textcircled{c} L = \frac{1}{2} \|X\beta - y\|_2^2 \rightarrow \nabla L(\beta) = 2(X^T X \beta - X^T y)$$

$$\mu = 0.1, \beta_0 = (1, 1)^T \rightarrow \text{error} = L(\beta_0) = 108.16$$

$$\nabla L = \begin{pmatrix} -201 \\ -157 \end{pmatrix}$$

$$\beta_1 = \beta_0 - \mu \nabla L = \begin{pmatrix} 21.1 \\ 16.7 \end{pmatrix} \rightarrow \text{error} = L(\beta_1) = 8812.7$$

$$\nabla L = \begin{pmatrix} 1818.9 \\ 1412.25 \end{pmatrix}$$

$$\beta_2 = \beta_1 - \mu \nabla L = \begin{pmatrix} -160.79 \\ -124.52 \end{pmatrix} \rightarrow \text{error} = 718408.69 = L(\beta_2)$$