Assignment 1

Linear Regression

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Problem 1

Find the closed form solution for linear regression that minimizes the mean square error (MSE)..

Solution. For Linear Regression loss function MSE, we have:

$$\begin{split} L(w) &= \frac{1}{2} \sum_{i=1}^{n} (w^{T} x_{i} - y_{i})^{2} \\ &= \frac{1}{2} (Xw - y)^{T} (Xw - y) \\ &= \frac{1}{2} (w^{T} X^{T} Xw - y^{T} Xw - w^{T} X^{T} y + y^{T} y) \\ &= \frac{1}{2} (w^{T} X^{T} Xw - 2y^{T} Xw + y^{T} y) \end{split}$$

Finding the first-order partial derivative (gradient) of L(w):

$$\begin{split} \nabla_w L(w) &= \frac{\partial L(w)}{\partial w} \\ &= \frac{\partial}{\partial w} \big[\frac{1}{2} (w^T X^T X w - 2 w^T X^T y + y^T y) \big] \\ &= X^T X w - X^T y \\ &= X^T (X w - y) \end{split}$$

let $\nabla_w L(w) = 0$, we have:

$$\nabla_w L(w) = 0$$
$$X^T X w - X^T y = 0$$

Finding the secound-order partial derivative (gradient) of L(w):

$$\begin{split} H(L(w)) &= \frac{\partial (\nabla X^T(Xw - y))}{\partial w} \\ &= X^T X \end{split}$$

Because $X^TX \geq 0$ so that, when $\nabla_w L(w) = 0$, w has its minimized value:

$$w^* = (X^T X)^{-1} X^T y$$