## **Functions of Linear Regression**

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Given a data set  $X \in \mathbb{R}^{n \times (m-1)}$  and a label set  $Y \in \mathbb{R}^{n \times 1}$ . Let two indices i and j be subject to  $1 \leq i \leq n$  and  $2 \leq j \leq m$  respectively. Thus, a data value is denoted as  $X^i_j$  and a label value is denoted as  $Y^i$ . For all  $X^i$ , insert  $X^i_1 = 1$  so that  $X \in \mathbb{R}^{n \times m}$  and  $1 \leq j \leq m$ . Initialize weight matrix  $W \in \mathbb{R}^{1 \times m}$  and thus a weight value is denoted as  $W_j$ .

Hypothesis function is defined in Equation (1).

$$H(X^{i}) = \sum_{j=1}^{m} W_{j} X_{j}^{i}$$
 (1)

Loss function is defined in Equation (2).

$$C(X^i) = H(X^i) - Y^i \tag{2}$$

Cost function is defined in Equation (3).

$$J(W) = \frac{1}{2n} \sum_{i=1}^{n} C^{2}(X^{i})$$

$$= \frac{1}{2n} \sum_{i=1}^{n} (\sum_{j=1}^{m} W_{j} X_{j}^{i} - Y^{i})^{2}$$
(3)

The partial derivative of J(W) with respect to  $W_j$  is given as follows:

$$\begin{split} \frac{\partial J(W)}{\partial W_j} &= \frac{1}{2n} \sum_{i=1}^n \frac{\partial C^2(X^i)}{\partial W_j} \\ &= \frac{1}{2n} \sum_{i=1}^n \frac{\partial C^2(X^i)}{\partial C(X^i)} \cdot \frac{\partial C(X^i)}{\partial H(X^i)} \cdot \frac{\partial H(X^i)}{\partial W_j} \\ &= \frac{1}{2n} \sum_{i=1}^n 2C(X^i) \cdot 1 \cdot X_j^i \\ &= \frac{1}{n} \sum_{i=1}^n (H(X^i) - Y^i) X_j^i \end{split}$$

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Thus, for all  $W_j$ , the update function is given as follows:

$$W_{j} = W_{j} - \frac{\gamma}{n} \sum_{i=1}^{n} (H(X^{i}) - Y^{i}) X_{j}^{i}$$

where  $\gamma$  is the learning rate given by users.