

Linear regression model

$$f_{w,b}(x) = wx + b$$

Cost function

$$J(w, b) = \frac{1}{2m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})^2$$

Gradient descent algorithm

repeat until convergence {

$$w = w - \alpha \frac{\partial}{\partial w} J(w, b) \rightarrow \frac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)}) x^{(i)}$$

$$b = b - \alpha \frac{\partial}{\partial b} J(w, b) \rightarrow \frac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})$$

}

→ Expanding the derivatives term.

→ It is actually $\frac{\partial}{\partial w}$ of cost function.

→ Let's see the derivation.

(Optional)

$$\frac{\partial}{\partial w} J(w, b) = \frac{\partial}{\partial w} \frac{1}{2m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})^2 = \frac{\partial}{\partial w} \frac{1}{2m} \sum_{i=1}^m (\underline{wx^{(i)} + b} - y^{(i)})^2$$

$$= \frac{1}{2m} \sum_{i=1}^m (\underline{wx^{(i)} + b} - y^{(i)}) \cancel{2} x^{(i)} = \frac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)}) x^{(i)}$$

$$\frac{\partial}{\partial b} J(w, b) = \frac{\partial}{\partial b} \frac{1}{2m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})^2 = \frac{\partial}{\partial b} \frac{1}{2m} \sum_{i=1}^m (\underline{wx^{(i)} + b} - y^{(i)})^2$$

$$= \frac{1}{2m} \sum_{i=1}^m (\underline{wx^{(i)} + b} - y^{(i)}) \cancel{2} = \frac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})$$

no $x^{(i)}$

→ Detailed Formula.

Gradient descent algorithm

repeat until convergence {

$$w = w - \alpha \frac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)}) x^{(i)}$$
$$b = b - \alpha \frac{1}{m} \sum_{i=1}^m (f_{w,b}(x^{(i)}) - y^{(i)})$$

}

$\frac{\partial}{\partial w} J(w, b)$

$\frac{\partial}{\partial b} J(w, b)$

Update
w and b
simultaneously

$$f_{w,b}(x^{(i)}) = wx^{(i)} + b$$