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DeepLearning.AI



Training Linear Regression

Gradient Descent for Linear Regression

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 {

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

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

}

(Optional)

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

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

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

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

no $x^{(i)}$

Gradient descent algorithm

repeat until convergence {

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

}

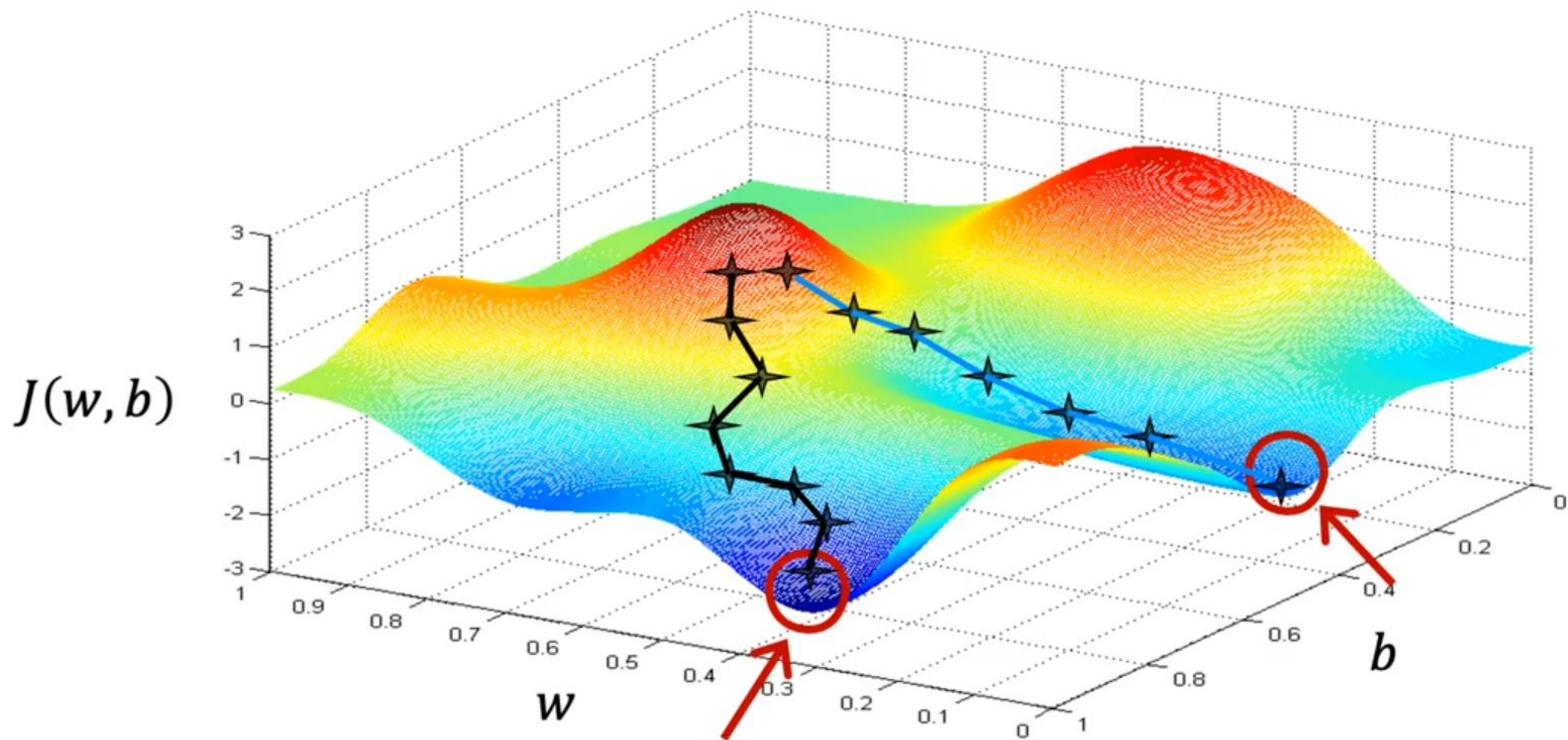
$\frac{\partial}{\partial w} J(w, b)$

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

Update w and b simultaneously

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

More than one local minimum



squared error cost

bowl shape 
convex function

