

Cost function: Squared error cost function

$$J(w, b) = \frac{1}{2m} \sum_{i=1}^m (\underbrace{\hat{y}^{(i)} - y^{(i)}}_{\text{error}})^2$$

m = number of training examples

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

intuition

Find w, b :

$\hat{y}^{(i)}$ is close to $y^{(i)}$ for all $(x^{(i)}, y^{(i)})$.

$f = wx + b$
 $= 9(200) + 0 = 1800$

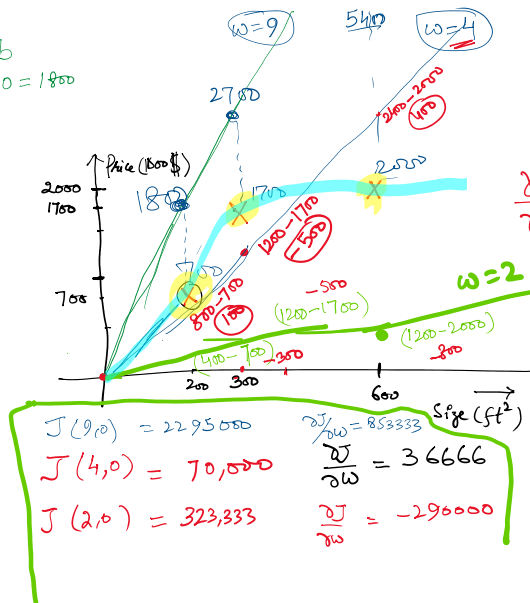
$w, b = 0$
 $w = 9$

prediction

$$J = \frac{1}{2m} \sum_{i=1}^3 (\hat{y}^i - y^i)^2$$

$$= \frac{1}{6} \left[(1800 - 700)^2 + (2700 - 1700)^2 + (5400 - 2000)^2 \right]$$

$$= 2295000$$



$$J = \frac{1}{2m} \sum (\hat{y} - y^i)^2$$

$$J = \frac{1}{2m} \sum (wx^i + b - y^i)^2$$

$$\frac{\partial J}{\partial w} = \frac{1}{2m} \sum [(wx^i + b - y^i) x^i]$$

$$= \frac{1}{3} [9(200) - 700]200 + \dots]$$

$$= 853333$$

$\frac{\partial J}{\partial w} = +ve$

$w_{new} = w_{old} \pm \Delta$

$w = w \pm \Delta$

$w = w - \alpha \frac{\partial J}{\partial w}$

$w = 9 - 5$

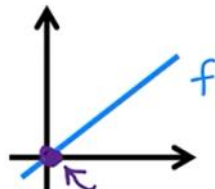
$w = 4$

simplified

$$f_w(x) = wx$$

w

$$b = 0$$



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

minimize $J(w)$

$$wx^{(i)}$$

model:

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

parameters:

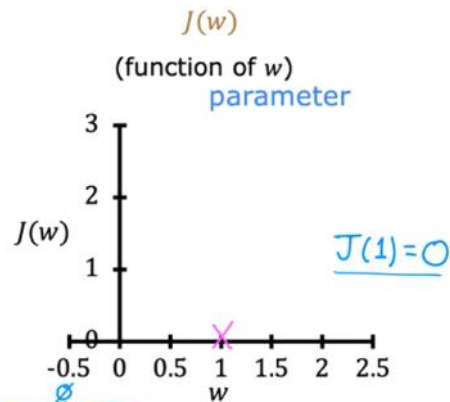
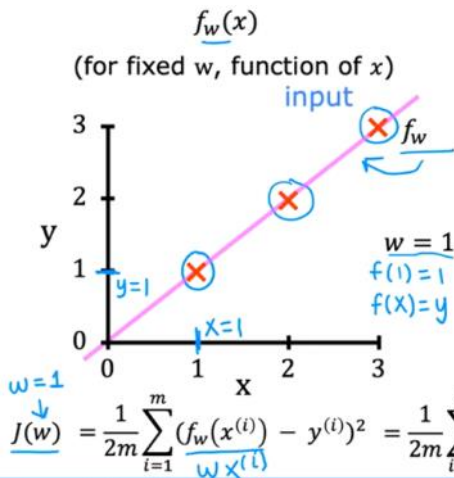
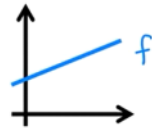
$$w, b$$

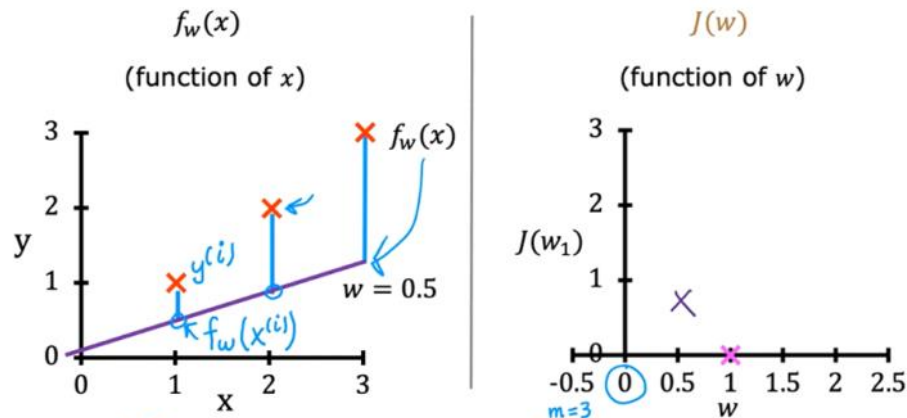
cost function:

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

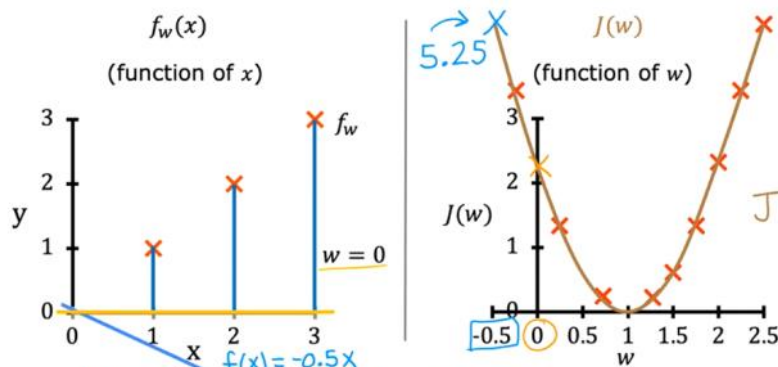
goal:

$$\text{minimize}_{w,b} J(w, b)$$

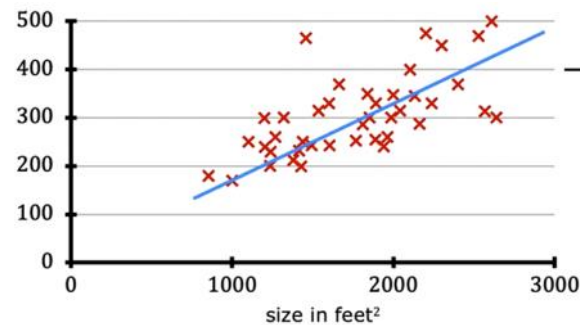




$$J(0.5) = \frac{1}{2m} [(0.5-1)^2 + (1-2)^2 + (1.5-3)^2] = \frac{1}{2 \cdot 3} [3.5] = \frac{3.5}{6} \approx 0.58$$



$$J(0) = \frac{1}{2m} (1^2 + 2^2 + 3^2) = \frac{1}{6} [14] \approx 2.3$$



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

... .. $w x^{(i)}$

goal of linear regression:

minimize $J(w)$

general case:

minimize $J(w, b)$