

Training set

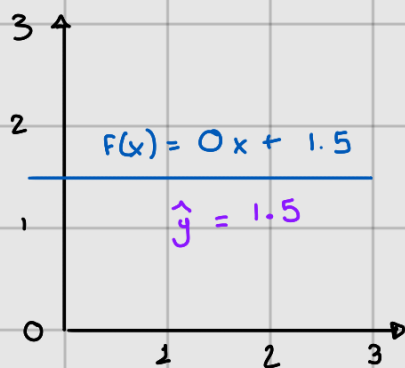
Features size in feet ² (x)	targets price \$1000's (y)
2000	400
1000	200
...	...

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

w, b : parameters

- coefficients
- weights

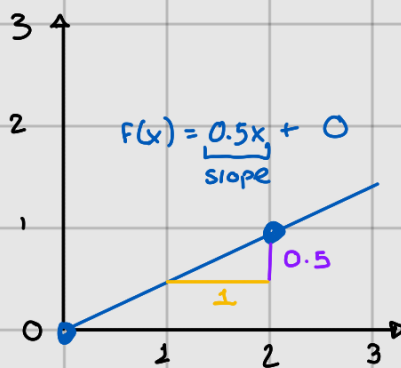
What do w, b do?



$w = 0$

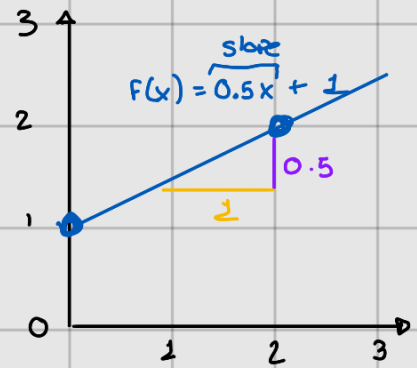
$b = 1.5$

↖ y-intercept



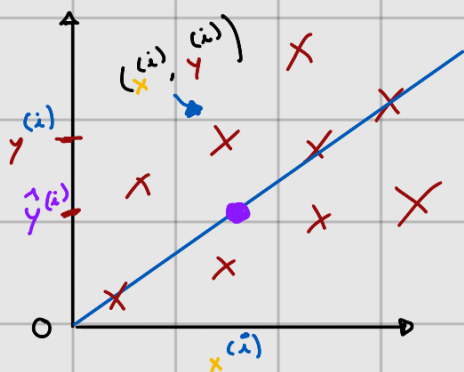
$w = 0.5$

$b = 0$



$w = 0.5$

$b = 1$



$\hat{y}^{(i)} = f_{w,b}(x^{(i)})$

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

Find w, b

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

Cost Function : Squared error cost function

$$J(w, b) = \frac{1}{2m} \sum_{i=1}^m \left(\overset{\text{Error}}{\hat{y}^{(i)} - y^{(i)}} \right)^2$$

m = number of training examples

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

Cost Function Intuition

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

parameters: w, b

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

\rightarrow model's predictions
 \rightarrow true values

goal: We want to minimize $J(w, b)$

Simplified

If we get a $f_w(x) = wx$ $b = 0$

w

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

minimize $J(w)$

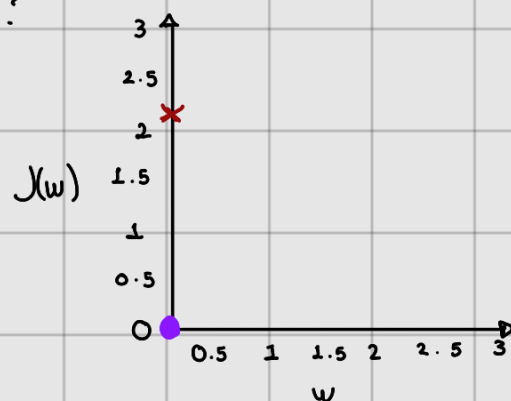
Activity

What is the cost function when w is 0?

$$J(0) = \frac{1}{2(3)} \left[\cancel{(0(1) - 1)^2} + \cancel{(0(2) - 2)^2} + \cancel{(0(3) - 2)^2} \right] =$$

$$\frac{1}{2(3)} \left[\frac{(-1)^2}{1} + \frac{(-2)^2}{4} + \frac{(-3)^2}{9} \right] =$$

$$\frac{1}{2(3)} [1 + 1 + 1] = \frac{1}{2(3)} [3] = \frac{1}{2} = 0.5$$



goal of linear regression: minimize $J(w)$

general case: minimize $J(w, b)$
 w, b

when the cost is relatively small, it means the model fits the data better compared to other choices for w and b