

Machine Learning

## Linear regression with one variable

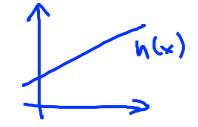
# Cost function intuition I

#### Hypothesis:

$$h_{\theta}(x) = \theta_0 + \theta_1 x$$

#### Parameters:

$$\theta_0, \theta_1$$



#### **Cost Function:**

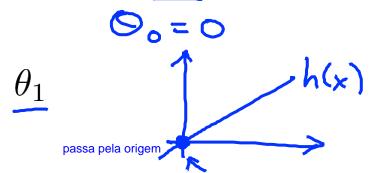
$$J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^{m} \left( h_{\theta}(x^{(i)}) - y^{(i)} \right)^2$$

Goal: minimize  $J(\theta_0, \theta_1)$ 

### Simplified

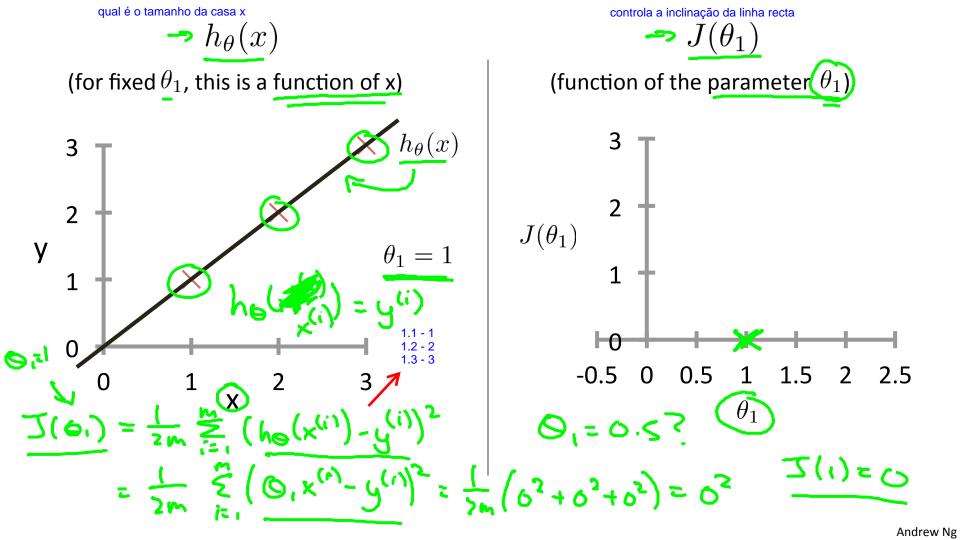
para percebermos formula anterior vamos SIMPLIFICAR imaginemos que x0=0 e x1=x

$$h_{\theta}(x) = \underline{\theta_1} x$$



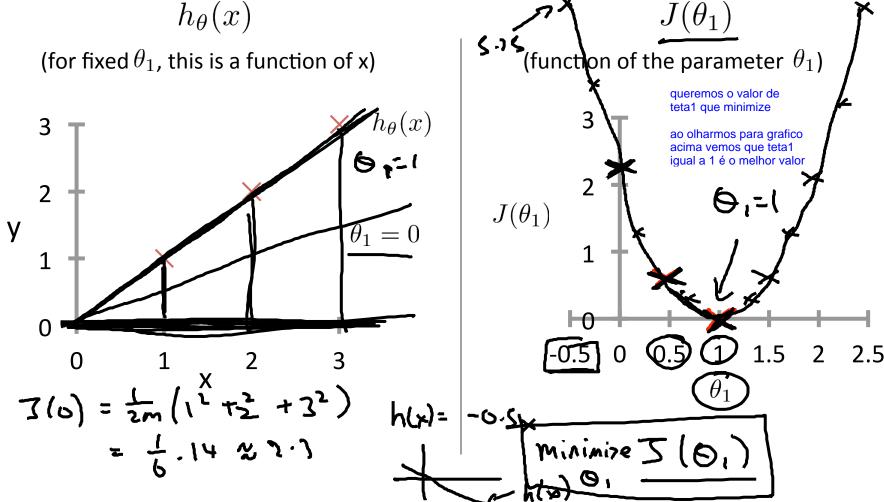
$$J(\theta_1) = \frac{1}{2m} \sum_{i=1}^{m} \left( h_{\theta}(x^{(i)}) - y^{(i)} \right)^2$$

$$\underset{\theta_1}{\text{minimize}} J(\theta_1) \qquad \diamondsuit_{\prime} \checkmark^{(\prime)}$$



$$h_{\theta}(x)$$
 (for fixed  $\theta_1$ , this is a function of x) (function of the parameter  $\theta_1$ ) 
$$\frac{3}{2}$$
 
$$y = \frac{1}{2} \sum_{k=1}^{\infty} \left[ (0.5 - 1)^k + (1 - 2)^k + (1 - 2)^k + (1 - 2)^k \right] = \frac{1}{2 \times 3} \left[ (3.5) = \frac{3.5}{2 \times 3} \times 0.105 \right]$$

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