LECTURE 3

# HOW TO MINIMIZE COST

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## Acknowledgement

- 01. Andrew Ng's ML Class
  - · https://class.coursera.org/ml-003/lecture
  - http://www.holehouse.org/mlclass(note)
- 02. Convolutional Neural Networks for Visual Recognition
  - · http://cs231n.github.io
- 03. Tensorflow
  - · https://www.tensorflow.org
  - · https://github.com/aymericdamien/TensorFlow-Examples

## **Hypothesis and Cost**

$$H(x) = W + b$$

cost(W,b) = 
$$\frac{1}{m} \sum_{i=1}^{m} (H(x^{(i)}) - y^{(i)})^2$$

## Simplified Hypothesis

$$H(x) = Wx$$

$$cost(W) = \frac{1}{m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)})^2$$

cost(W) = 
$$\frac{1}{m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)})^2$$

X	Y
1	1
2	2
3	3

W=1,cost(W)=?

$$cost(W) = \frac{1}{m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)})^2$$

X	Y
1	1
2	2
3	3

$$W=1,cost(W)=0$$

$$\frac{1}{3}((1x1-1)^2+(1x2-2)^2+(1x3-3)^2)$$

$$W=0,cost(W)=4.67$$

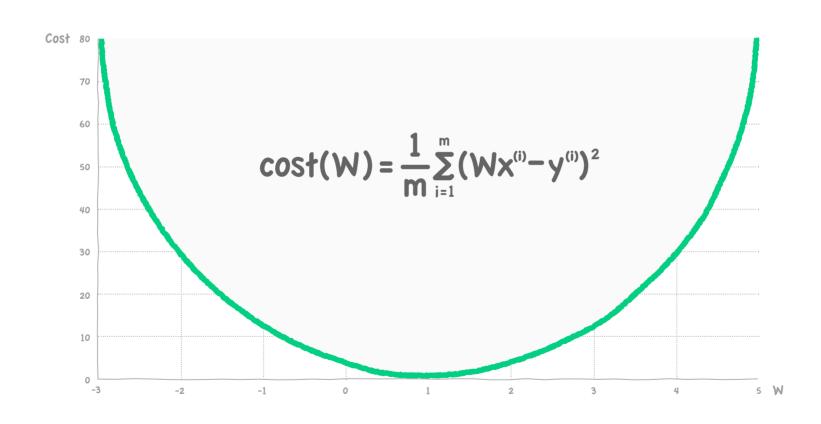
$$\frac{1}{3}((0x1-1)^2+(0x2-2)^2+(0x3-3)^2)$$

$$W=2,cost(W)=?$$

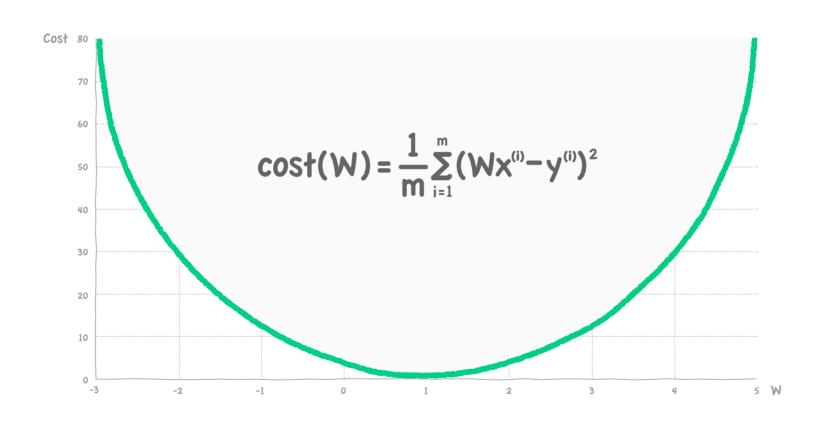
$$\cdot$$
 W=1, cost(W)=0

$$W=0$$
, cost(W)=4.67

$$W=2$$
,  $cost(W)=4.67$ 



## **How to Minimize Cost?**



## **Gradient Descent Algorithm**

- · Minimize cost function
- · Gradient descent is used many minimization problems
- · For a given cost function, cost(W, b), it will find W,b to minimize cost
- · It can be applied to more general function: cost(w1, w2, ...)

## **How It Works?**



#### **How It Works?**

- 01. Start with initial guesses
  - · Start at 0,0 (or any other value)
  - \* Keeping changing W and b a little bit to try and reduce cost(tW,b)
- 02. Each time you change the parameters, you select the gradient which reduces cost (W, b) the most possible
- 03. Repeat
- 04. Do so until you converge to a local minimum
- 05. Has an interesting property
  - · Where you start can determine which minimum you end up



#### **Formal Definition**

$$cost(W) = \frac{1}{m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)})^2$$



$$cost(W) = \frac{1}{2m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)})^2$$

#### **Formal Definition**

$$cost(W) = \frac{1}{2m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)})^2$$

$$W := W - \alpha \frac{\partial}{\partial W} \cos t(W)$$

#### **Formal Definition**

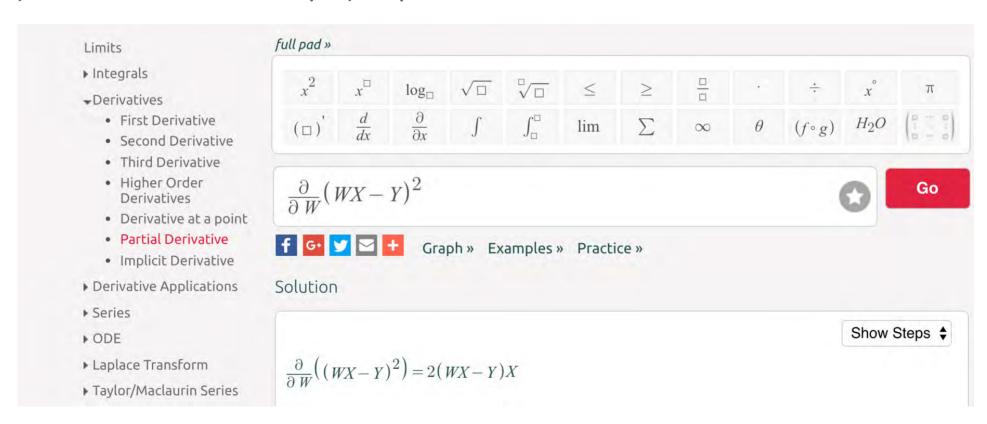
W: = W - 
$$\alpha \frac{\partial}{\partial W} \frac{1}{2m} \sum_{i=1}^{m} (W x^{(i)} - y^{(i)})^2$$

$$W := W - \alpha \frac{1}{2m} \sum_{i=1}^{m} 2(Wx^{(i)} - y^{(i)}) x^{(i)}$$

$$W := W - \alpha \frac{1}{m} \sum_{i=1}^{m} (W x^{(i)} - y^{(i)}) x^{(i)}$$

## **Partial Derivative Calculator**

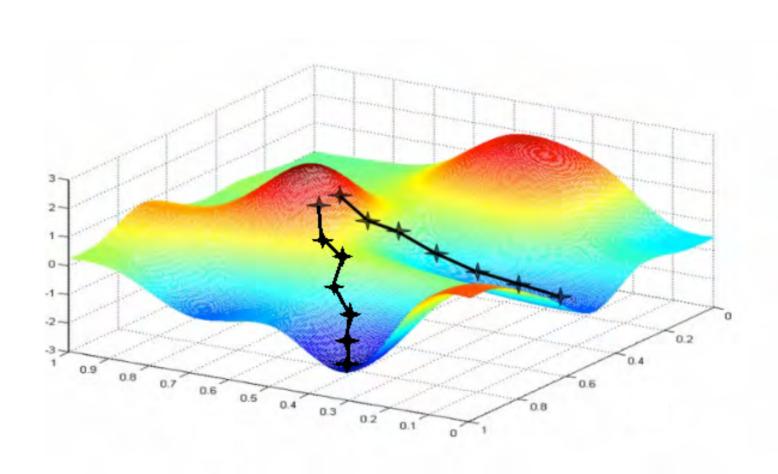
#### Partially differentiate functions step-by-step



## **Gradient Descent Algorithm**

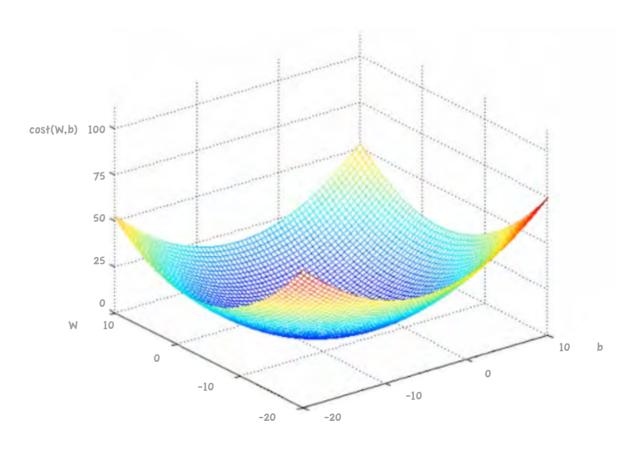
$$W := W - \alpha \frac{1}{m} \sum_{i=1}^{m} (Wx^{(i)} - y^{(i)}) x^{(i)}$$

## **Convex Function**



#### **Convex Function**

cost(W,b) = 
$$\frac{1}{m} \sum_{i=1}^{m} (H(x^{(i)}) - y^{(i)})^2$$



#### **NEXT LECTURE**

# MULTIVARIABLE LOGISTIC REGRESSION