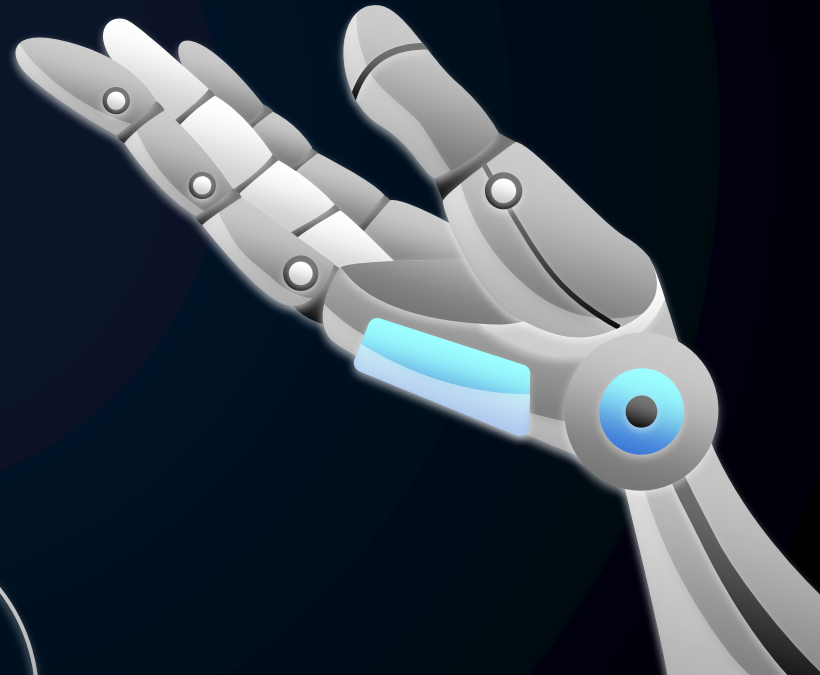
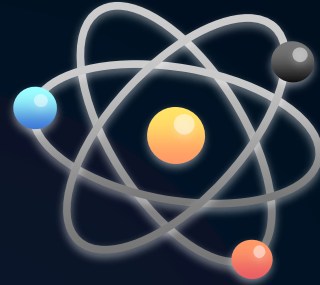
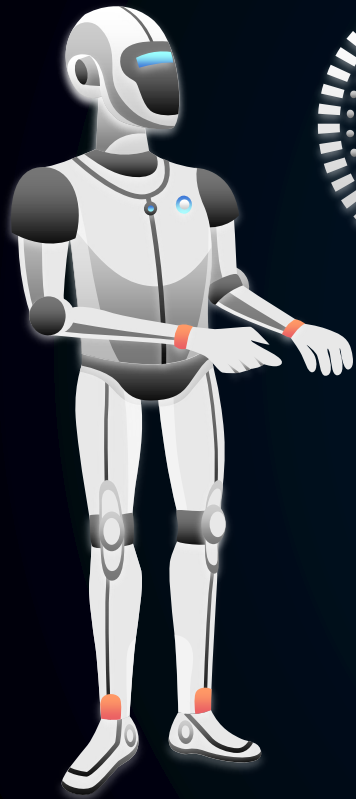


Introduction To Machine Learning

Waleed Alsarhani

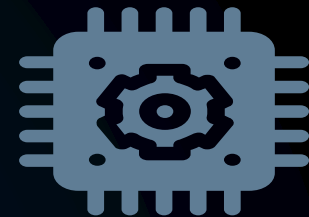
Eyad Alqaysi





Introduction





Agenda

1

AI

Definition of AI,
examples and false
information about it

2

Machine Learning

Definition,
examples, idea,
disadvantages and
advantages

3

Machine Learning Algorithms

Types, definitions,
models and
examples

4

Linear Regression

Definition, types,
examples,
equations, concept
of Optimization and
Gradient Descent

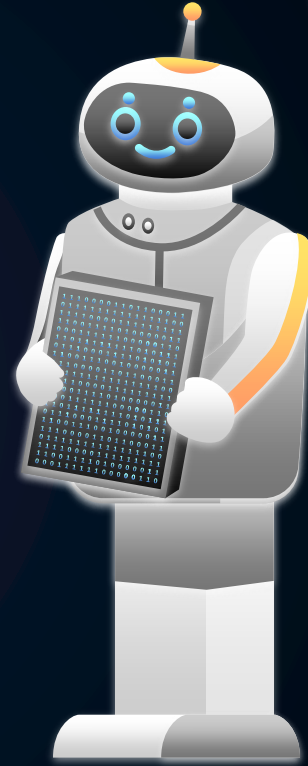


1

AI

What is AI ?

AI is the Systems or devices that simulate human intelligence to perform tasks.



Examples Of AI



mail

Message classification process



Social media

View content based on interests

False Information

1

Artificial intelligence
delivers magical results
instantly

AI needs time, money, goal
and planning in order to
output the excellent results

2

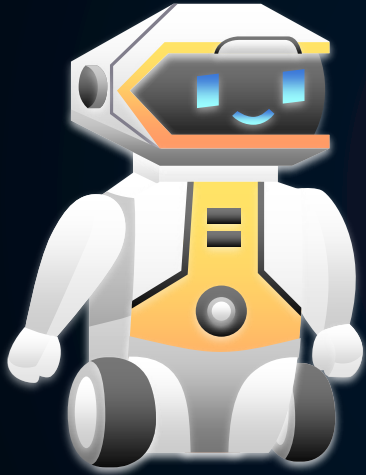
the more data the
better

Enterprise AI needs smart
data

3

Artificial intelligence
does not need humans

AI is not about bots taking
over



2

Machine Learning

What is Machine Learning ?

Machine learning is a branch of AI that is concerned with the design and development of algorithms and techniques that allow computers to have the "learning" advantage.



Examples Of ML



Healthcare

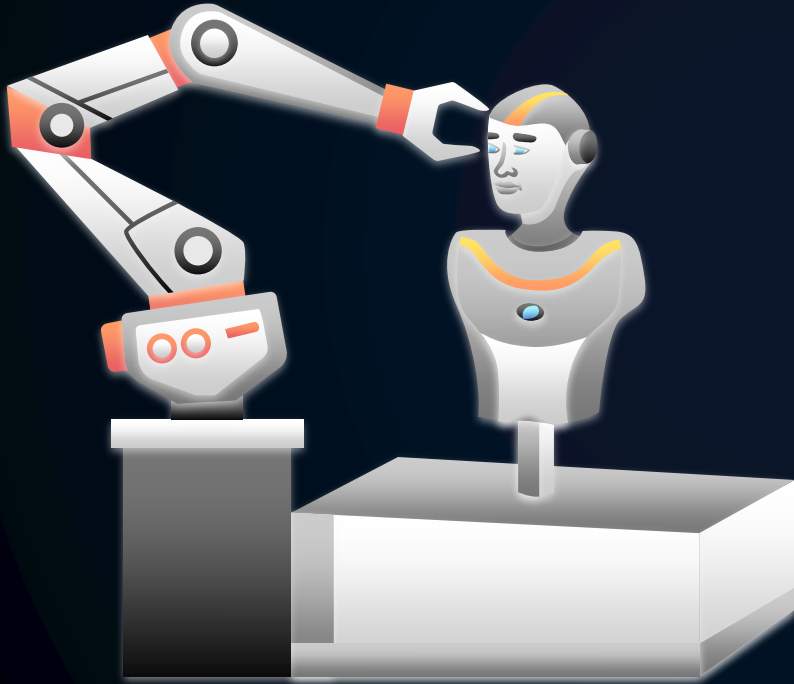
Detection of cancerous tumors



Retail sales

Stock management and sales
increase

Idea Of Machine Learning



Child and Machine

Advantages

1

Work without human
intervention after setup

2

More accurate results

3

Dealing with different
data of different sizes

Disadvantages

1

Expensive and time
consuming

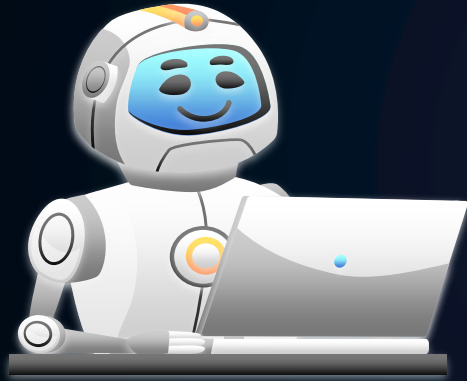
2

It requires a large initial
investment if the
devices are set up on
premises

3

It is difficult to correctly
interpret the results
without the help of
experts

3



Machine Learning Algorithms

What is Machine Learning Algorithms?

A machine learning algorithm is the method by which the AI system performs its task



Types of machine learning algorithms

1

Supervised Learning

Give the machine the
inputs and outputs

2

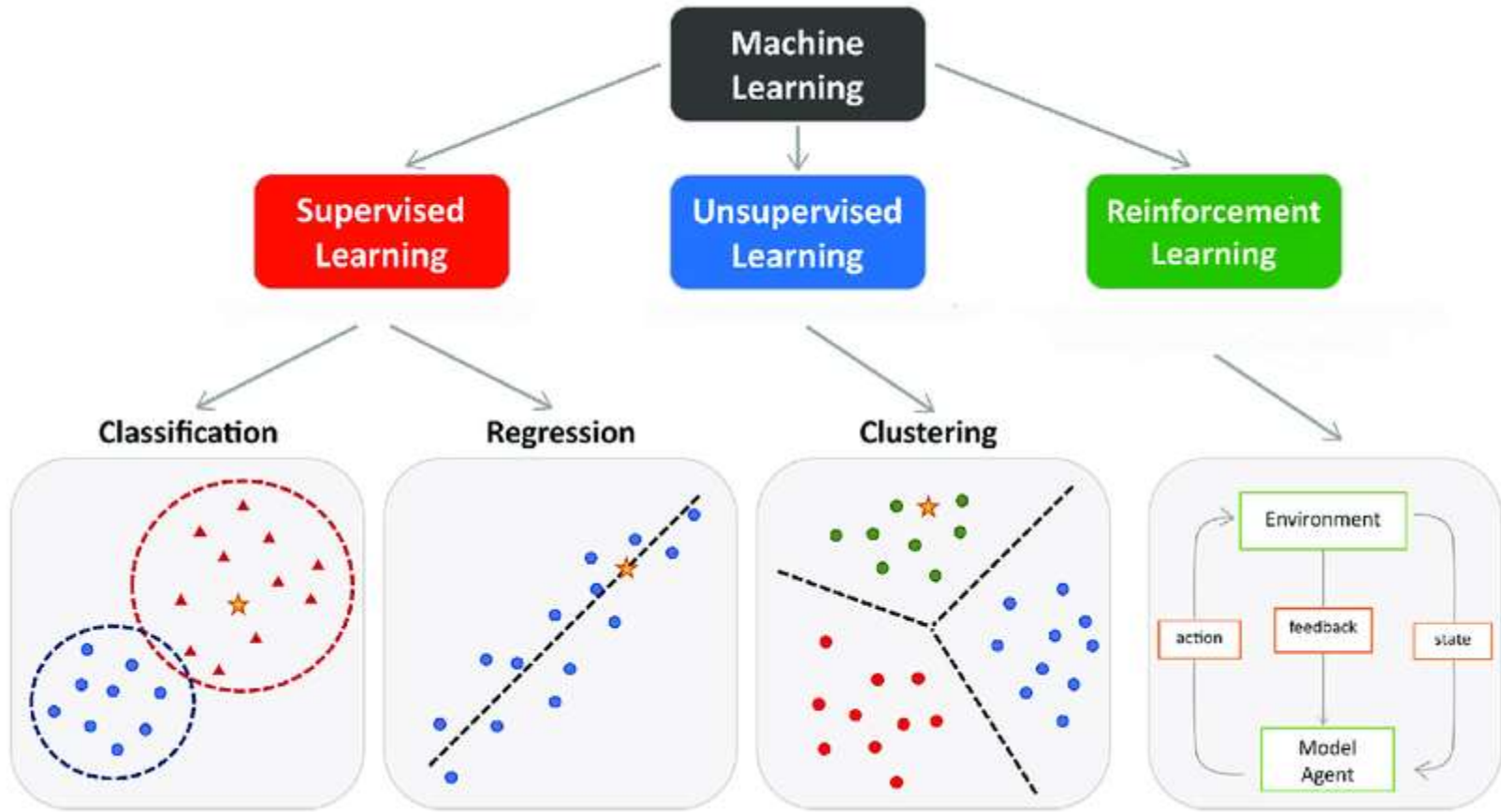
Unsupervised Learning

Give the machine the
inputs

3

Reinforcement learning

principle of reward and
punishment



Types of machine learning algorithms

Classification

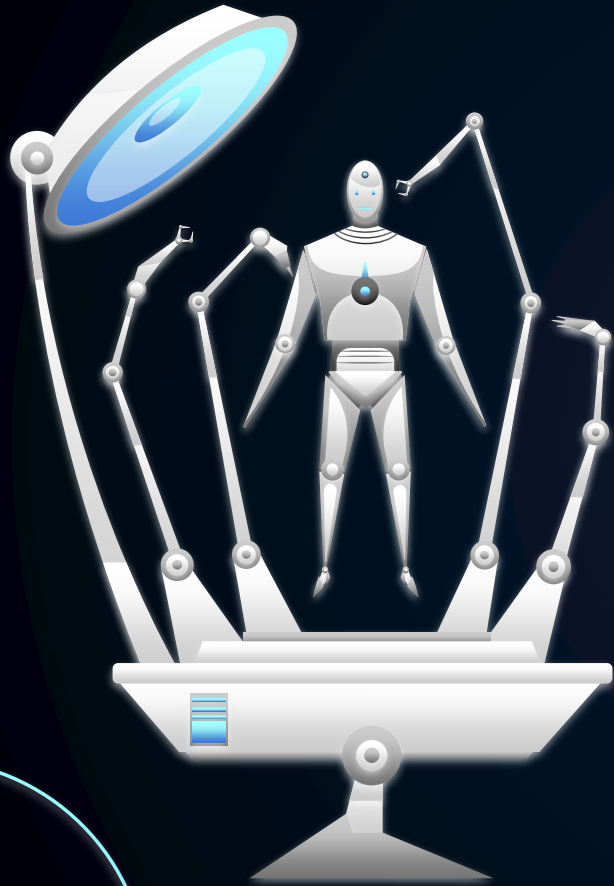
Determining whether the tumor is benign based on the medical image

Regression

Forecasting your company's revenue next year, based on many performance metric

Clustering

Identifying topics in a set of blog posts



4

Linear Regression

Linear regression equation

Linear regression equation
(without error)

$$\hat{Y} = bX + a$$

predicted
values of Y

b = slope = rate of
predicted \uparrow/\downarrow for Y
scores for each unit
increase in X

Y -intercept =
level of Y
when X is 0

Slope

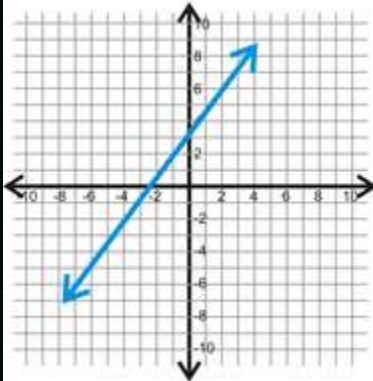
$$y_2 = 1 \quad y_1 = -7$$

$$m = \frac{y_2 - y_1}{x_2 - x_1} = \frac{1 - (-7)}{12 - (-4)}$$

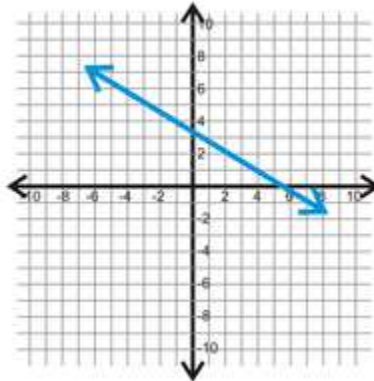
$$x_2 = 12 \quad x_1 = -4$$

Slope

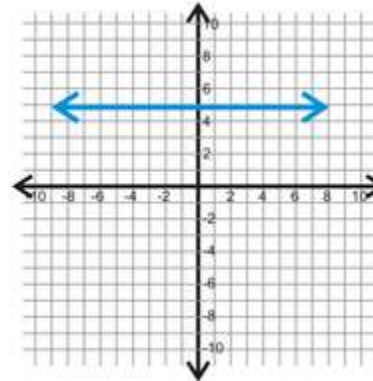
Positive



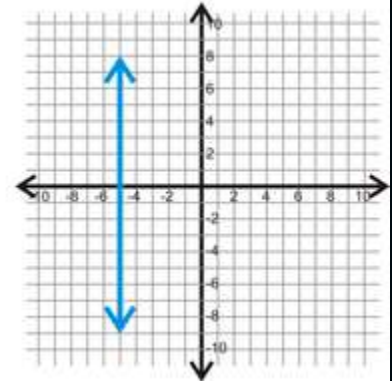
Negative



Zero

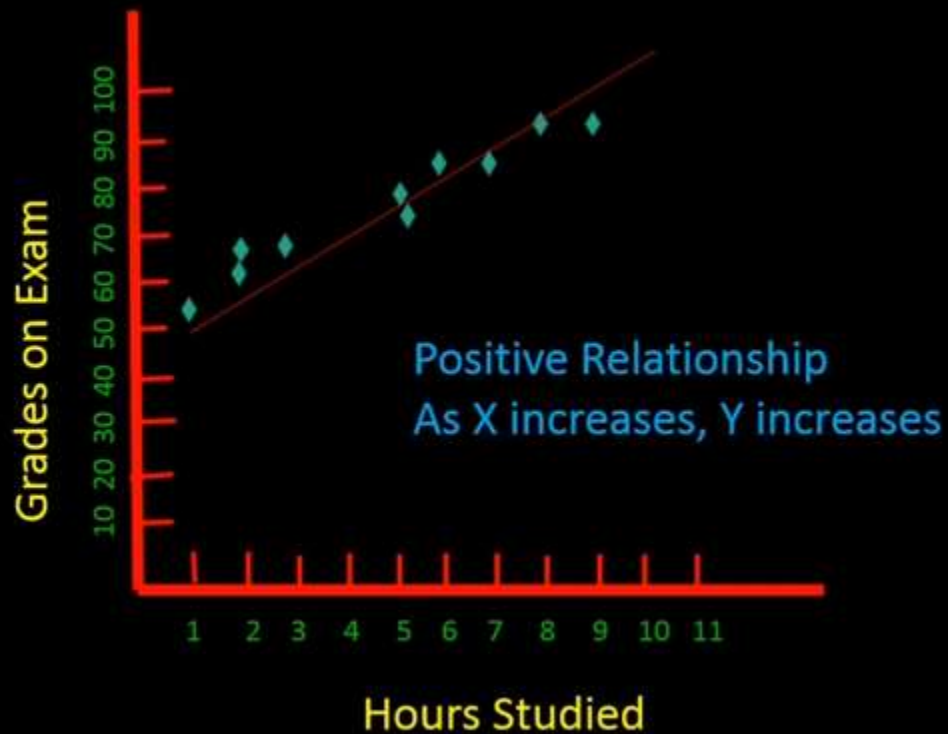


Undefined



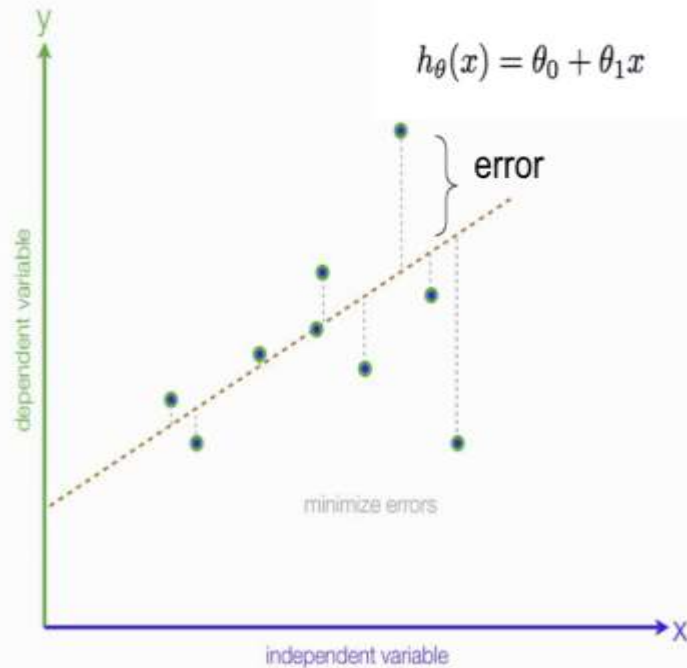
Linear regression

Hours Studied	Grade on Exam
2.00	69.00
9.00	98.00
5.00	82.00
5.00	77.00
3.00	71.00
7.00	84.00
1.00	55.00
8.00	94.00
6.00	84.00
2.00	64.00



Input X	Inputs
Output Y	Outputs
Rows m	Rows
Features n	Elements
$h(X)$	Predicted Value
Cost J	error value
Theta	X Transactions

Cost Function



Hypothesis:

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

Parameters:

$$\theta_0, \theta_1$$

Cost Function:

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

Goal:

$$\underset{\theta_0, \theta_1}{\text{minimize}} J(\theta_0, \theta_1)$$

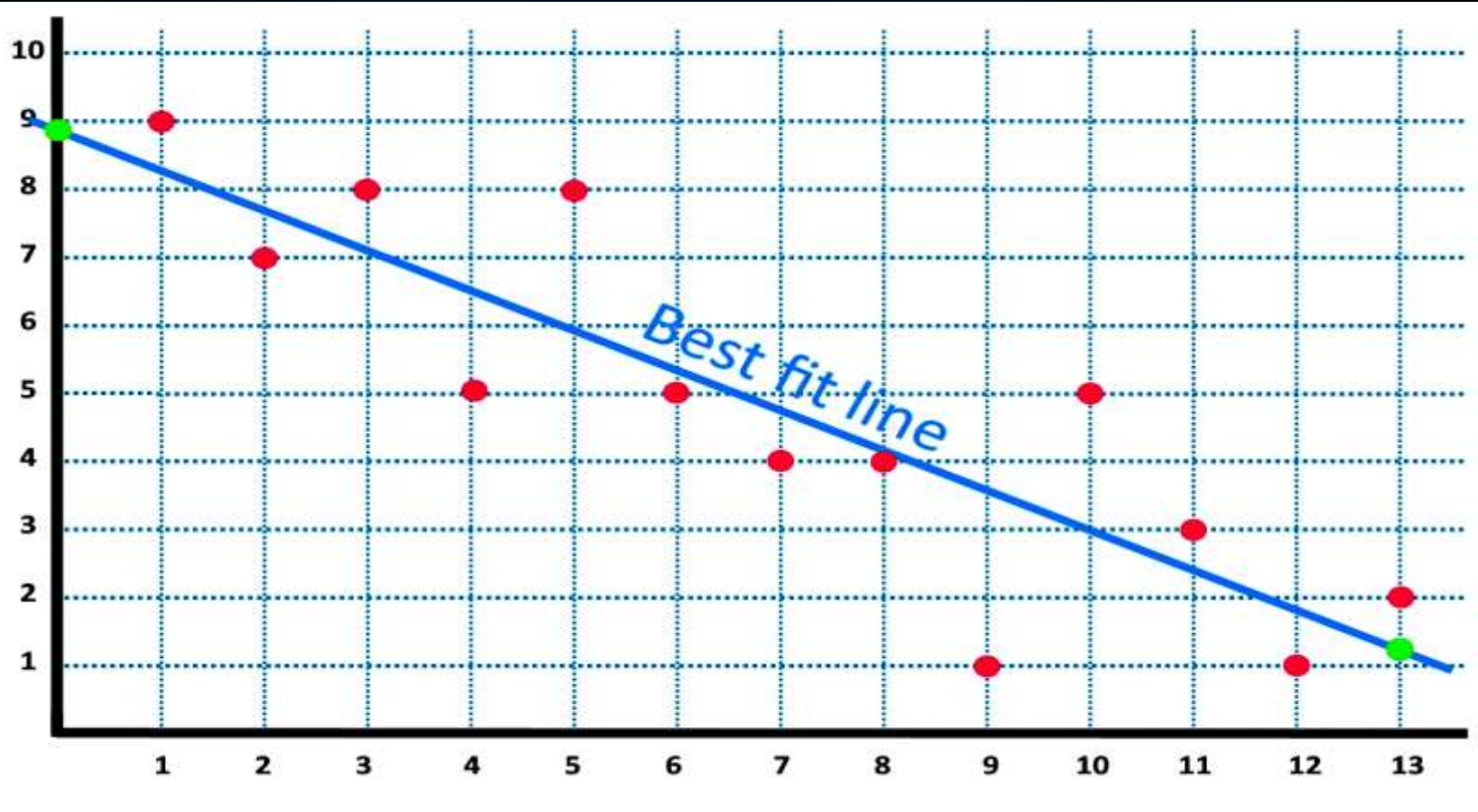
Theta 0 =5

Theta 1 =2

$h(x) = 5 + 2x$

X	Y	$h(X)$	$h(X)-y$	$(h(X)-y)^2$
1	7			
2	8			
2	7			

Best Fit Line



Optimization

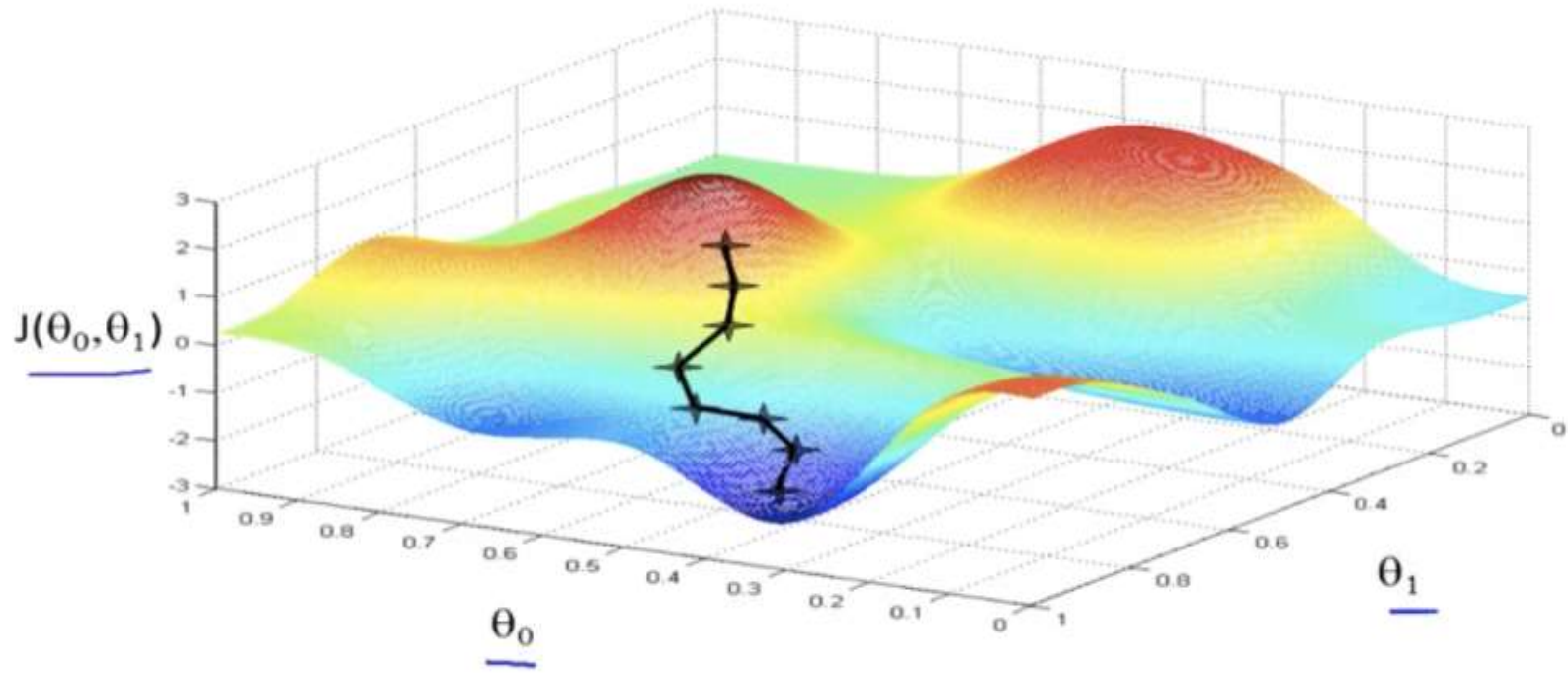
aims to search for the best choices according to (constraints) to get the best results

(maximization) or (minimization).

Gradient Descent

It is an optimization algorithm that aims to reach the lowest point in theta so that the cost function is the lowest

Gradient Descent



Gradient Descent

$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$$

Gradient Descent

repeat until convergence {

$$\theta_0 := \theta_0 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})$$

$$\theta_1 := \theta_1 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \cdot x^{(i)}$$

}

Gradient Descent

Correct: Simultaneous update

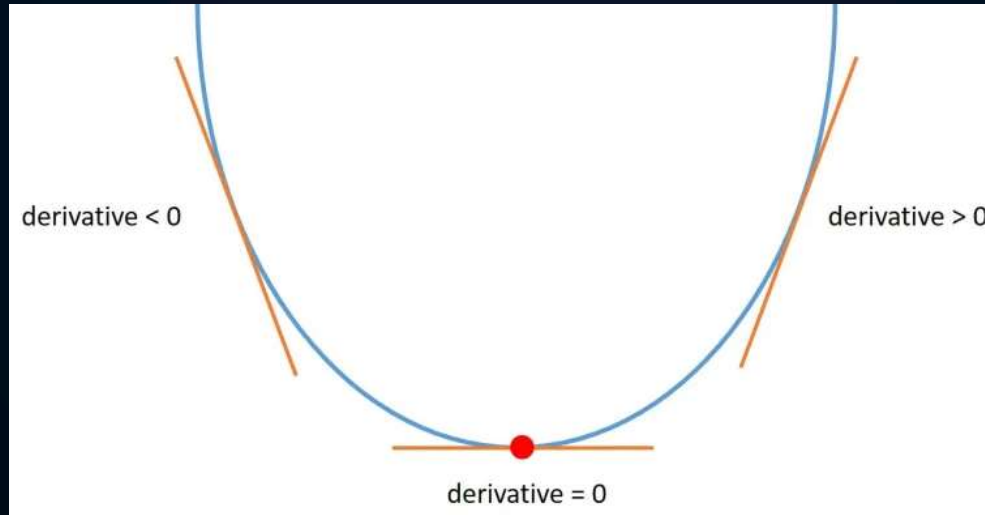
- $\text{temp0} := \theta_0 - \alpha \frac{\partial}{\partial \theta_0} J(\theta_0, \theta_1)$
- $\text{temp1} := \theta_1 - \alpha \frac{\partial}{\partial \theta_1} J(\theta_0, \theta_1)$
- $\theta_0 := \text{temp0}$
- $\theta_1 := \text{temp1}$

Incorrect:

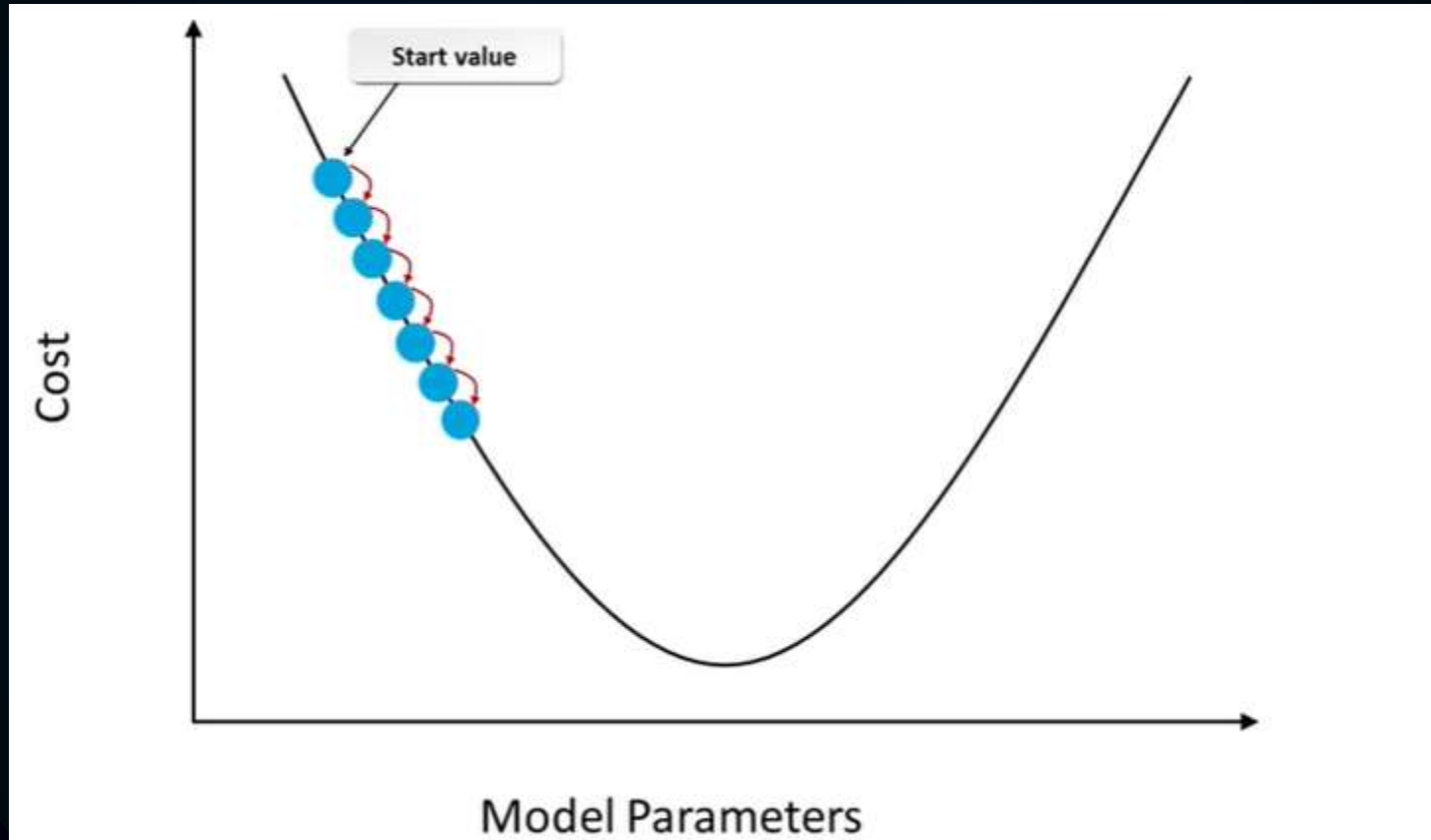
- $\text{temp0} := \theta_0 - \alpha \frac{\partial}{\partial \theta_0} J(\theta_0, \theta_1)$
- $\theta_0 := \text{temp0}$
- $\text{temp1} := \theta_1 - \alpha \frac{\partial}{\partial \theta_1} J(\theta_0, \theta_1)$
- $\theta_1 := \text{temp1}$

Gradient Descent

In the Gradient Descent algorithm, the direction will be directed downward, so if the differential value is positive, then the direction will be to the left (as in the figure), if it is negative, then the direction will be to the right, and it will stop when zero is reached, which means that the bottom will be reached Because there is no change



Gradient Descent



Gradient Descent

$$\theta_0 := \theta_0 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x_i) - y_i)$$

Theta 0 =1

Theta 1 =3

Alpha=0.002

$h(x) = 1 + 3x$

Home	Price	$h(X)$	$h(X)-y$
100	300		
95	285		
90	270		