

# ZUJO

Session 1: Introduction to Machine Learning Part 1

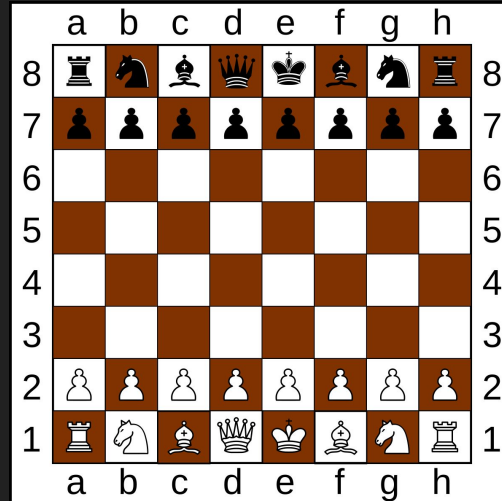
# Index

- Introduction to machine learning
- Comparing with human
- Understanding of supervised learning
- Linear Regression

# What is Machine Learning?

Machine learning is field of study that gives computers the ability to learn without being explicitly programmed.

- Arthur Samuel (1959)

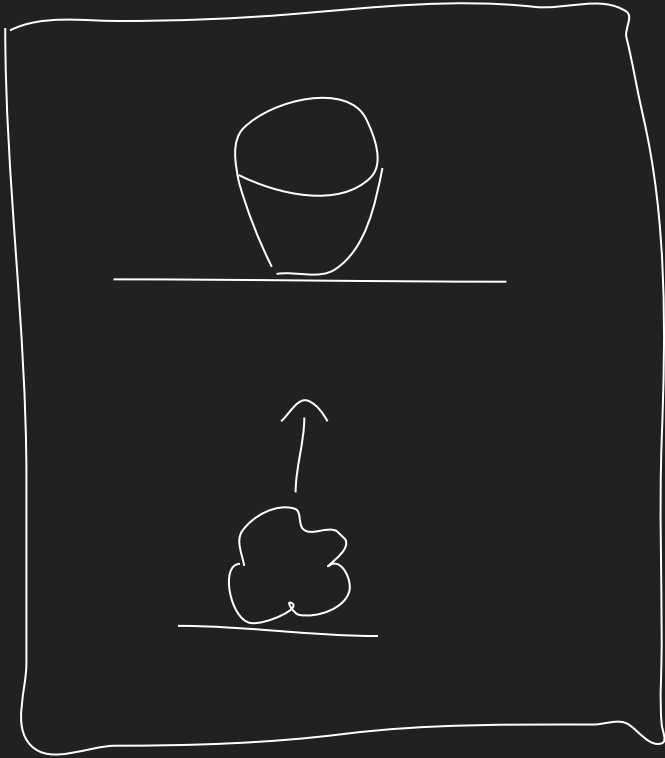


# What is Machine Learning?

A computer program is said to learn from experience  $E$  with respect to some class of tasks  $T$  and performance measure  $P$ , if its performance at tasks in  $T$ , as measured by  $P$ , improves with experience  $E$ .

- Tom Mitchell (1998)

# In simple terms



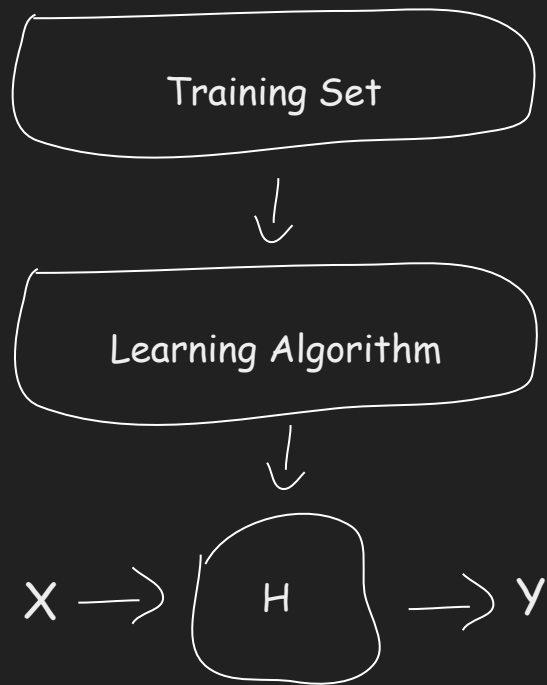
**1st Attempt:** You have to put too much force.

**2nd Attempt:** You realise you are closer to target but you need to increase your throw angle.

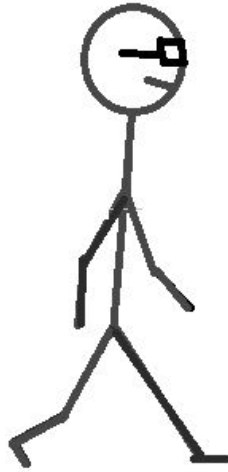
What is happening here is basically after every throw we are learning something and improving the end result.

We are programmed to learn from our experience.

# Learning Process



# Compare it with Humans



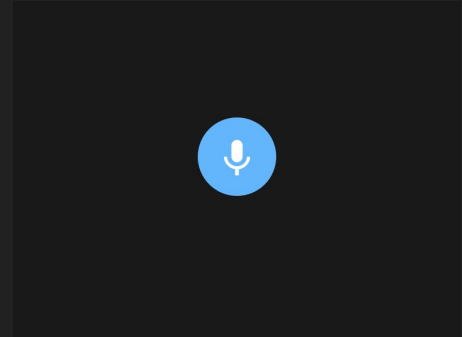
Human

Speak & Listen



AI

Statistical Learning  
Speech Recognition





Human

# Write and Read



AI

Statistical Learning

Speech  
Recognition

NLP

London is the capital and most populous city of England and the United Kingdom. Standing on the River Thames in the south east of the island of Great Britain, London has been a major settlement for two millennia. It was founded by the Romans, who named it Londinium. London's ancient core, the City of London, largely retains its 1.12-square-mile (2.9 km<sup>2</sup>) medieval boundaries.

Human

See



AI

Statistical Learning

Symbolic Learning

Speech Recognition

NLP

Computer Vision



Human

# Process and Visualise



AI

# Image Processing

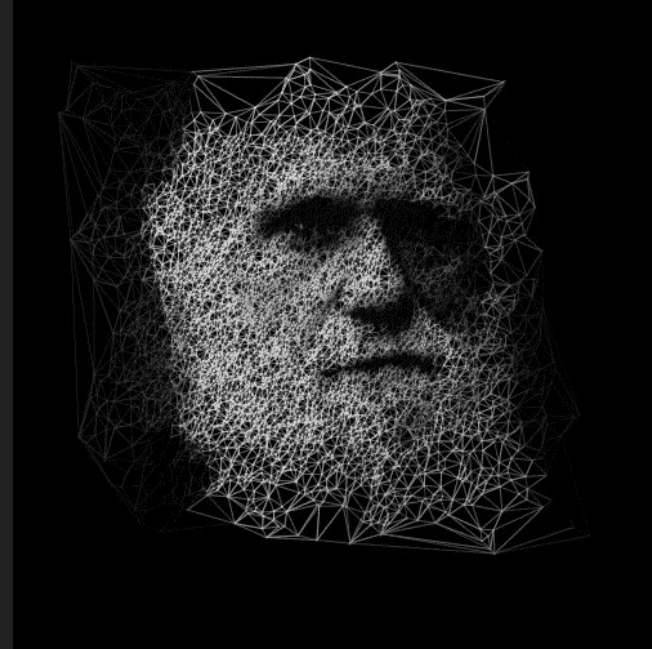
Statistical Learning

Speech Recognition

NLP

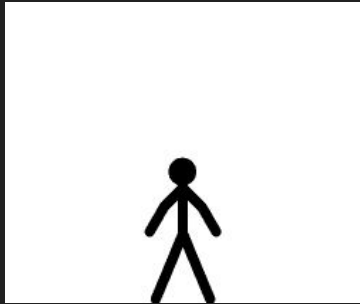
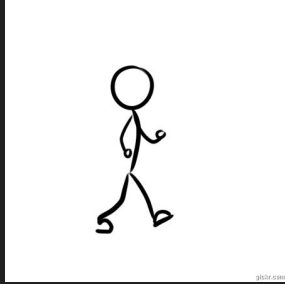
Symbolic Learning

Computer Vision



Human

# Understand Environment



AI

Image Processing

Statistical Learning

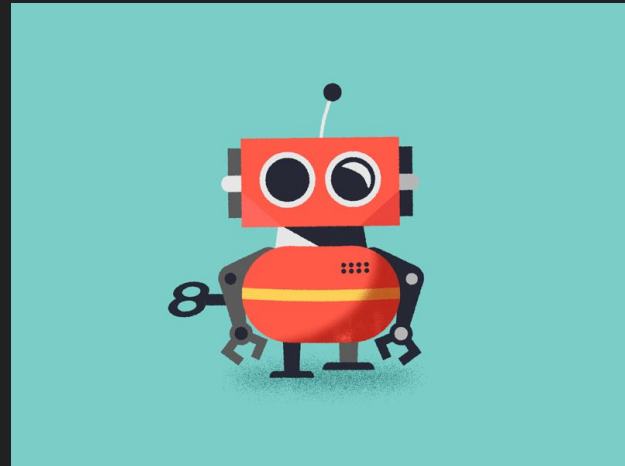
Symbolic Learning

Speech Recognition

NLP

Computer Vision

# Robotics



# Human Recognise Patterns



# Machine Learning

AI

Image Processing

Pattern  
Recognition

Symbolic Learning

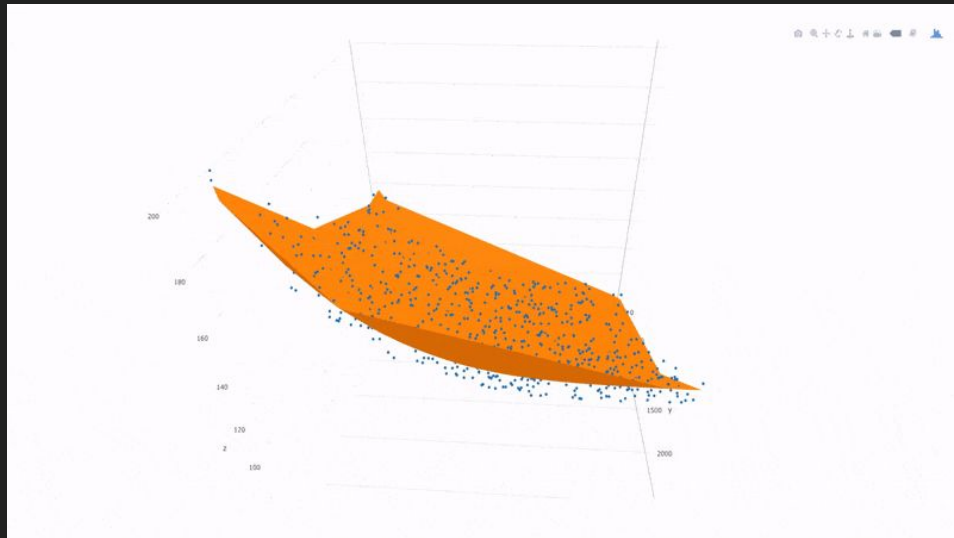
Robotics

Computer Vision

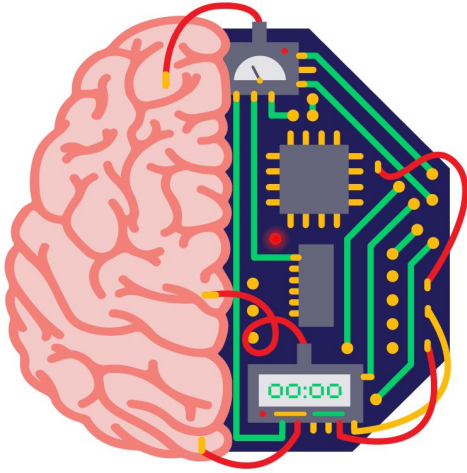
Statistical Learning

Speech Recognition

NLP



# Human Brain



AI

Pattern Recognition Machine Learning

Image Processing

Statistical Learning

Symbolic Learning

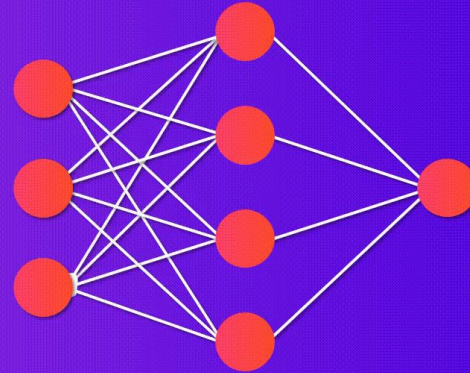
Speech Recognition NLP

Robotics

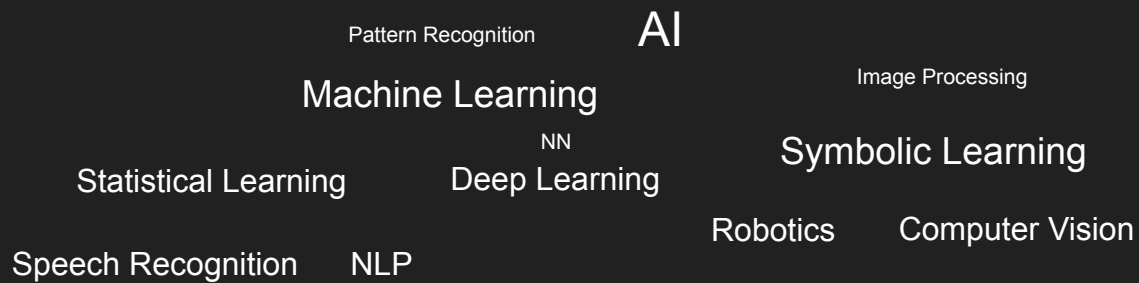
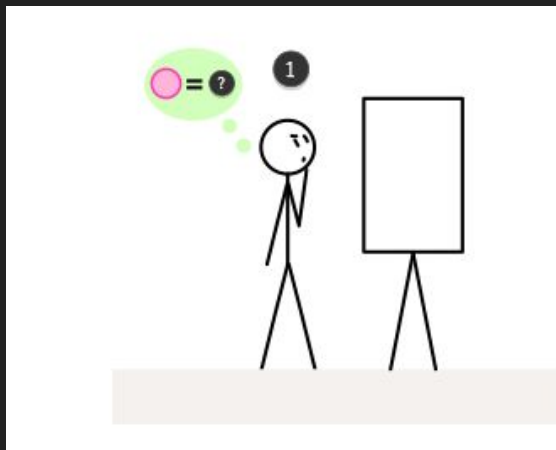
Computer Vision

# Deep Learning

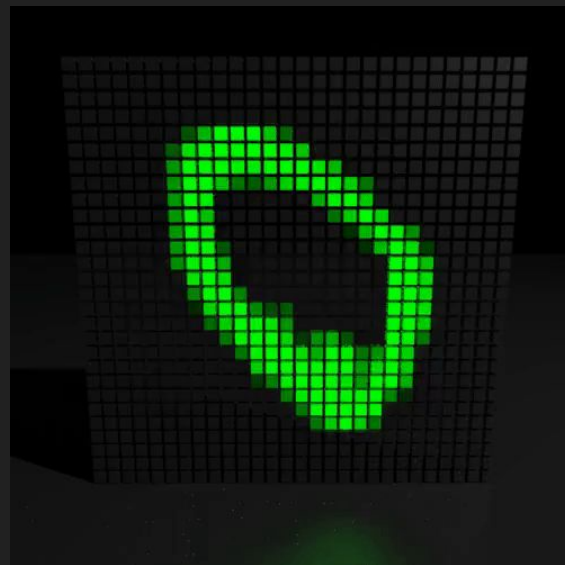
Neural Networks



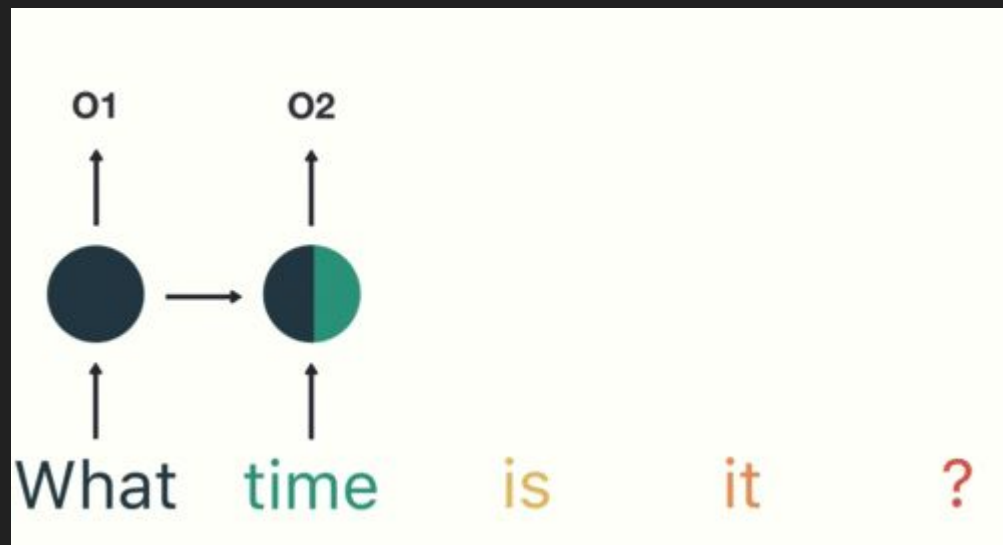
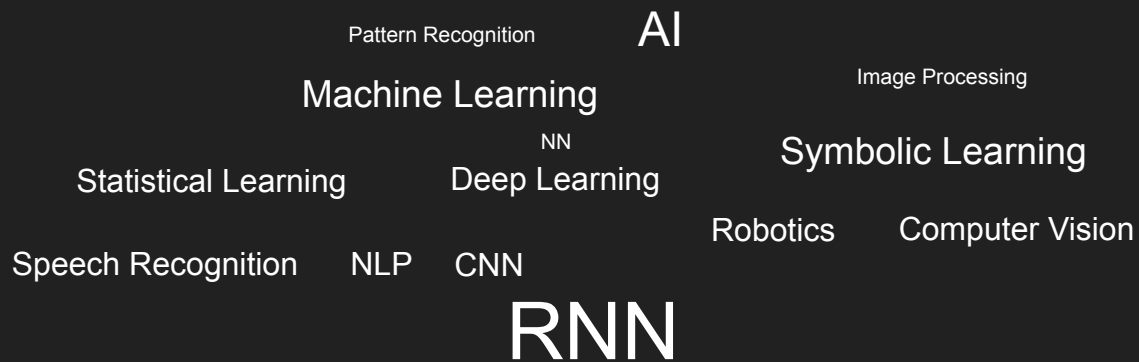
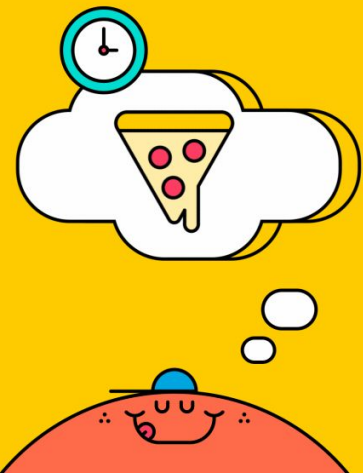
# Human Remember Scene



## CNN



# Human Remember Past





# Human Dream



Pattern Recognition

AI

Machine Learning

Image Processing

NN

Statistical Learning

Deep Learning

Symbolic Learning

Speech Recognition

NLP

CNN

RNN

Robotics

Computer Vision

# GAN



# AI

Pattern Recognition

Machine Learning

Statistical Learning

Deep Learning

Speech  
Recognition

NLP

CNN RNN GAN

Classification

Prediction

Image Processing

Symbolic Learning

Robotics

Computer Vision

Supervised Learning

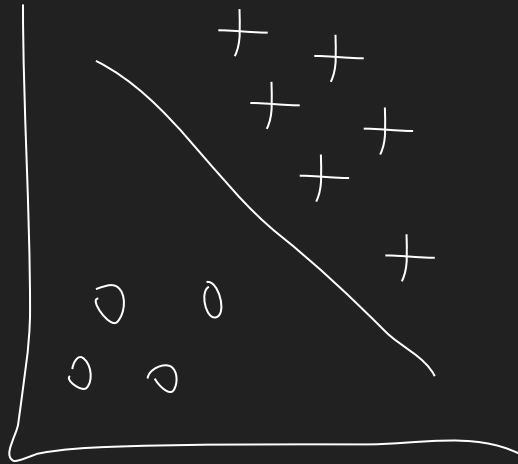
Unsupervised Learning

Reinforcement Learning

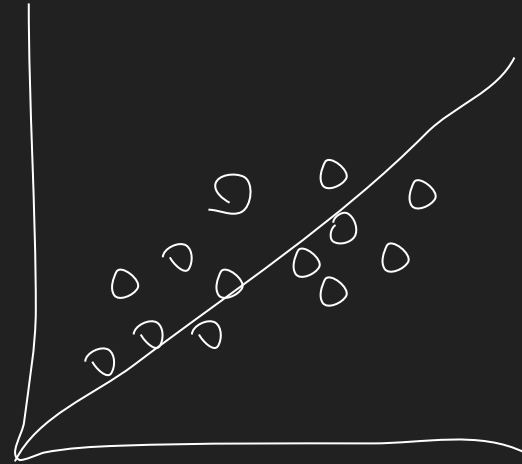
# Understanding each type

- Supervised Learning
- Unsupervised Learning
- Semi-supervised Learning
- Reinforcement Learning

# Supervised Learning

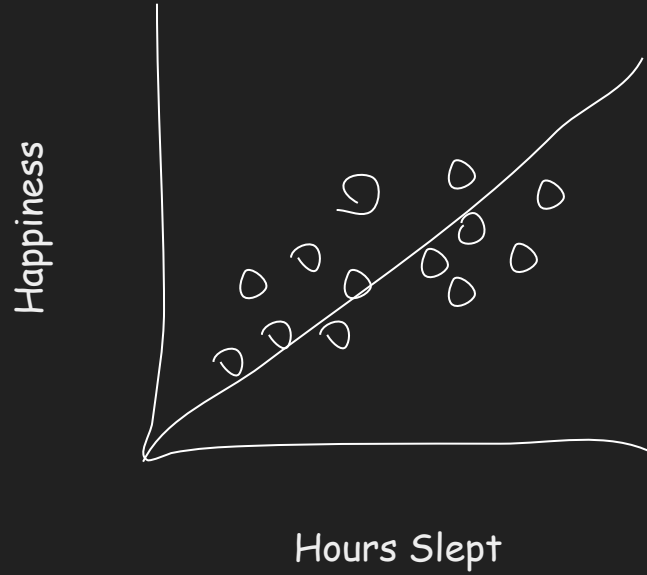


Classification



Regression

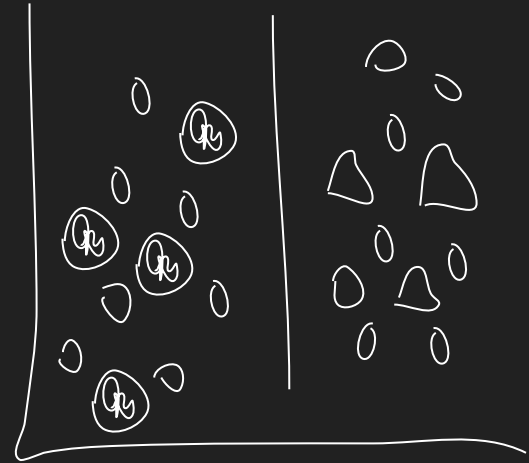
# Supervised Learning Example



# Semi-supervised Learning

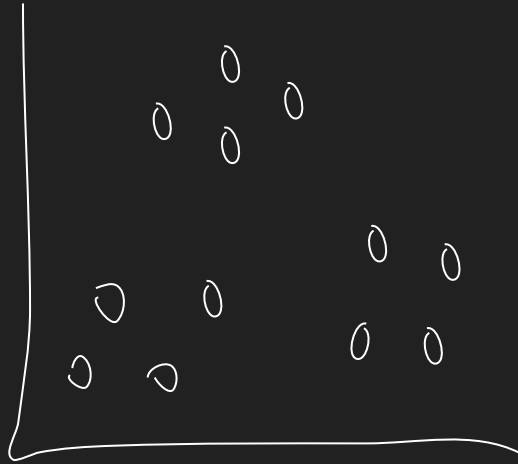


Labeled and Unlabeled  
data

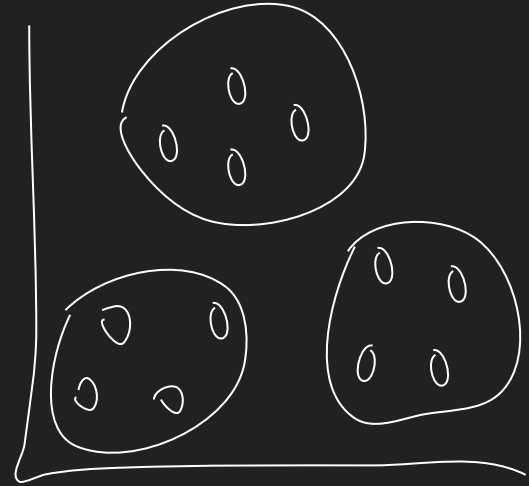


Semi-supervised Learning

# Unsupervised Learning



Before Clustering

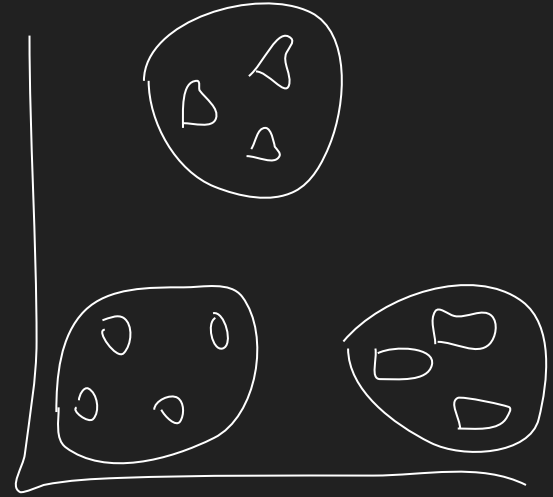


After Clustering

# Unsupervised Learning Example



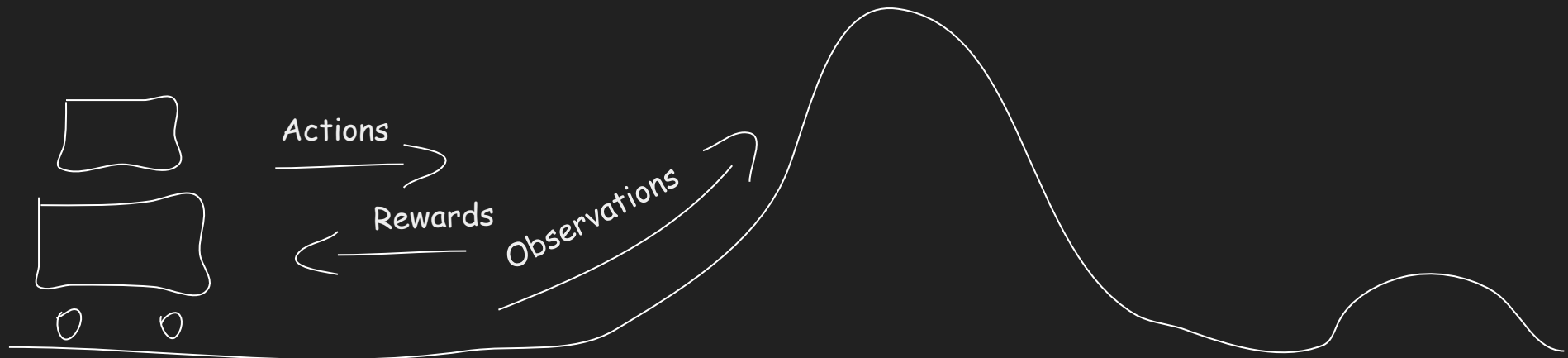
Before Clustering



After Clustering



# Reinforcement Learning



# Supervised Learning

- **Linear regression**
- Logistic regression
- Naive bayes
- **Support vector machine**
- Linear discriminant analysis
- Decision trees
- K-nearest neighbor
- **Neural networks (Multi-layer perceptron)**
- Similarity learning

# Unsupervised Learning Algorithms

- K-means clustering
- KNN (k-nearest neighbors)
- Hierarchical clustering
- Anomaly detection
- **Neural Networks**
- Principal Component Analysis
- Independent Component Analysis
- Apriori algorithm
- Singular value decomposition

# Semi-supervised Learning Algorithms

- Generative models
- Low-density separation
- Graph-based methods
- Heuristic approaches

# Reinforcement Learning Algorithms

- Monte Carlo
- Q-learning
- SARSA (State–action–reward–state–action)

# Linear Regression

# Dataset

Total Area	No of bedrooms	Prices (Lakh)
1056	2 BHK	39.07
2600	4 Bedroom	120
1440	3 BHK	62
1521	3 BHK	95
1200	2 BHK	51
1170	2 BHK	38
2732	4 BHK	204
3300	4 BHK	600

# Dataset





## Hypothesis

$$h(x) = \theta_0 + \theta x_1 + \theta x_2$$

$$x_1 = \textit{Size}$$

$$x_2 = \textit{Bedrooms}$$

# Hypothesis

$$h(x) = \sum_{j=1}^2 \theta_j x_j$$

Where,  $x_0 = 1$

$$\theta = \begin{bmatrix} \theta_0 \\ \theta_1 \\ \theta_2 \end{bmatrix} \quad x = \begin{bmatrix} x_0 \\ x_1 \\ x_2 \end{bmatrix}$$

$\theta = \text{parameters}$

$m = \# \text{ training examples}$

$x = \text{inputs}$

$y = \text{output}$

$(x, y) = \text{training example}$

$(x^{(i)}, y^{(i)}) = n^{\text{th}} \text{ training example}$

$n = \# \text{ features}$

How to choose theta?

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

Parameters  $\rightarrow (\theta_0 + \theta_1)$

Hypothesis

Simplified Math

Predicted Output

Actual Output

Row Index

Cost Function  $\rightarrow J(\theta)$

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

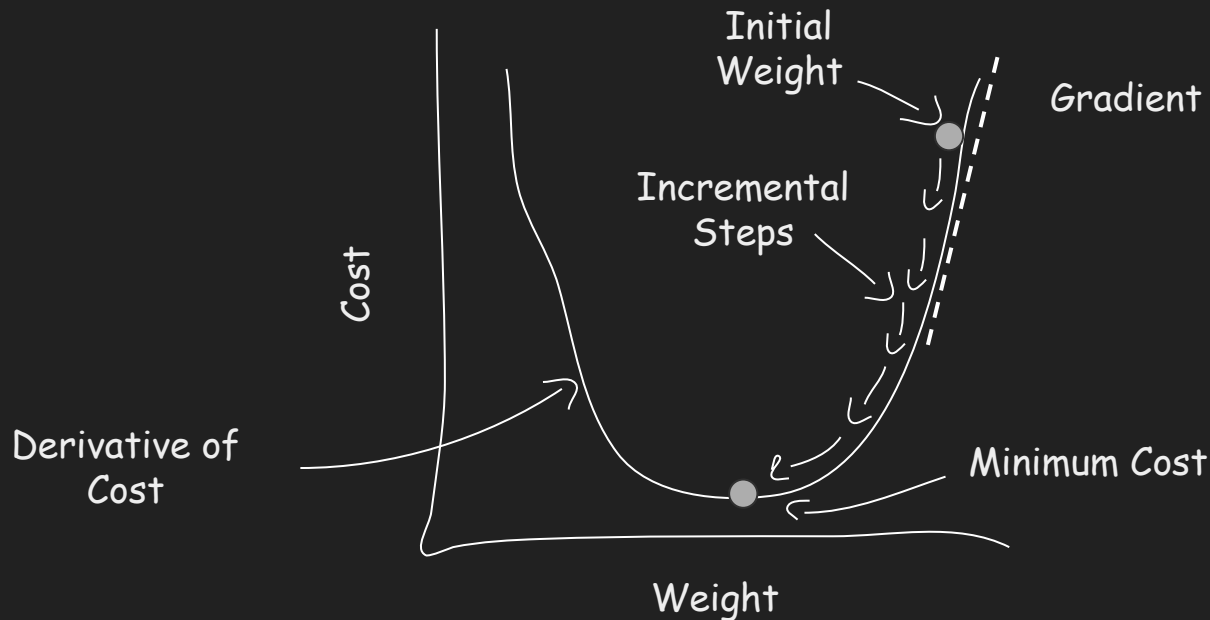
Summation of all the samples

Total Squared Error

Goal  $\rightarrow$  Minimize  $J(\theta)$

# Gradient Descent

Gradient descent is an iterative optimization algorithm for finding the local minimum of a function.



# Gradient Descent Algorithm

The goal of the gradient descent algorithm is to minimize the given function (say cost function). To achieve this goal, it performs two steps iteratively:

1. Compute the gradient (slope), the first order derivative of the function at that point
2. Make a step (move) in the direction opposite to the gradient, opposite direction of slope increase from the current point by  $\alpha$  times the gradient at that point

# Gradient Descent Algorithm

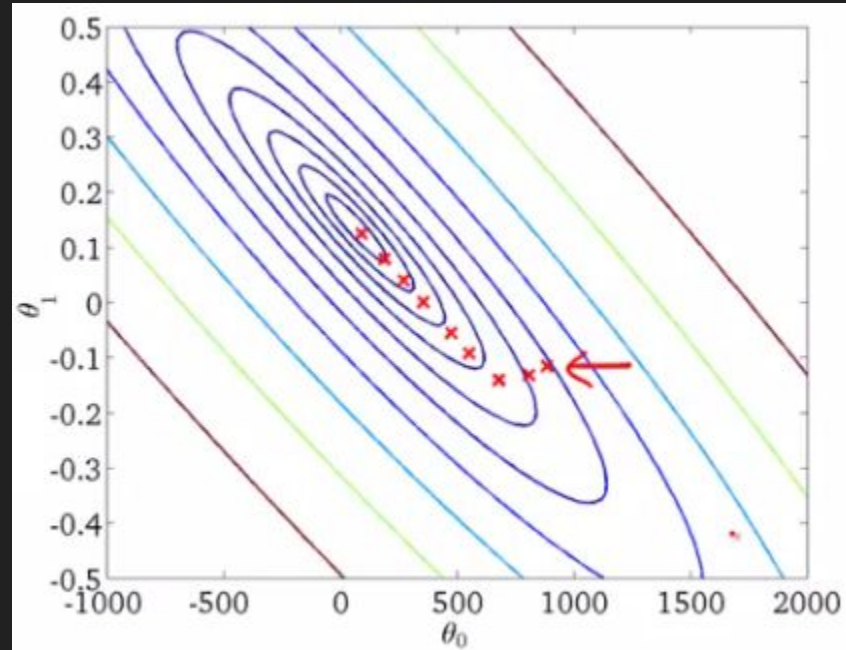
*repeat until convergence*{

$$\theta_j := \theta_j - \underbrace{\alpha}_{\text{Learning Rate}} \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$$

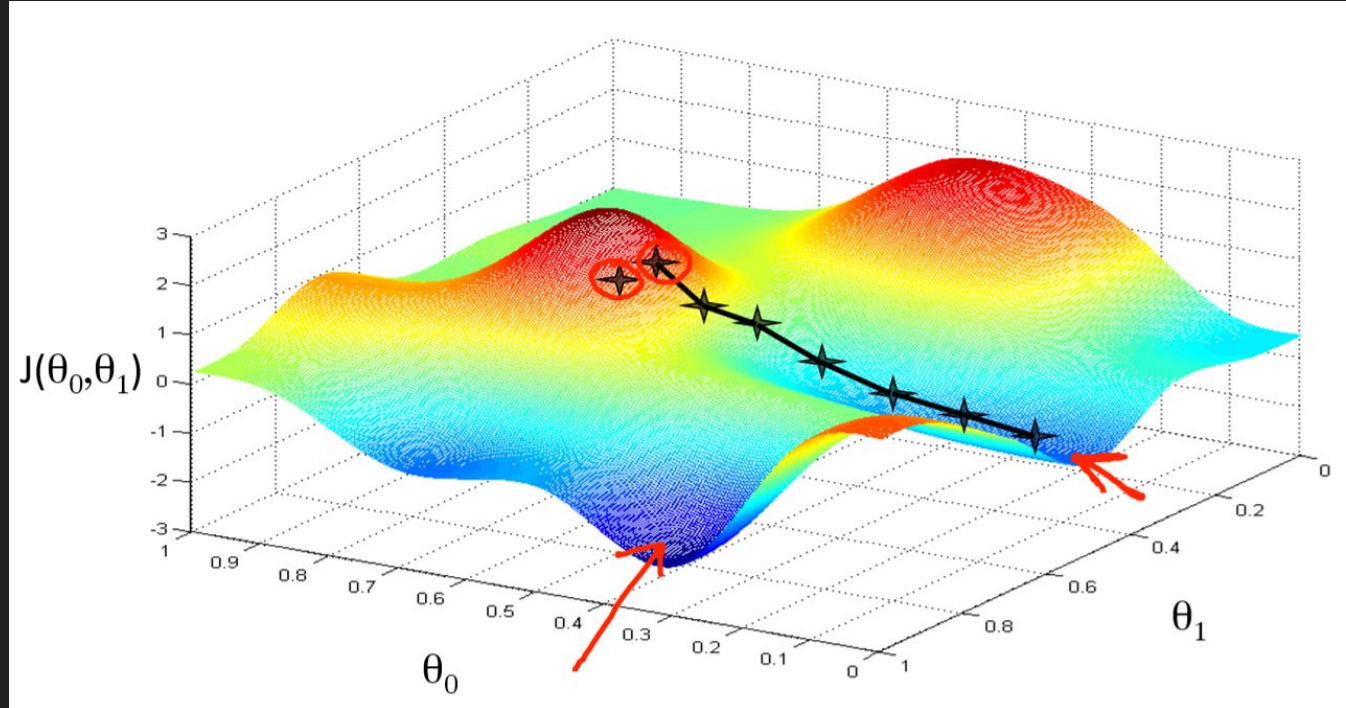
*(for  $j = 1$  and  $j = 0$ )*

}

# Gradient Descent Algorithm

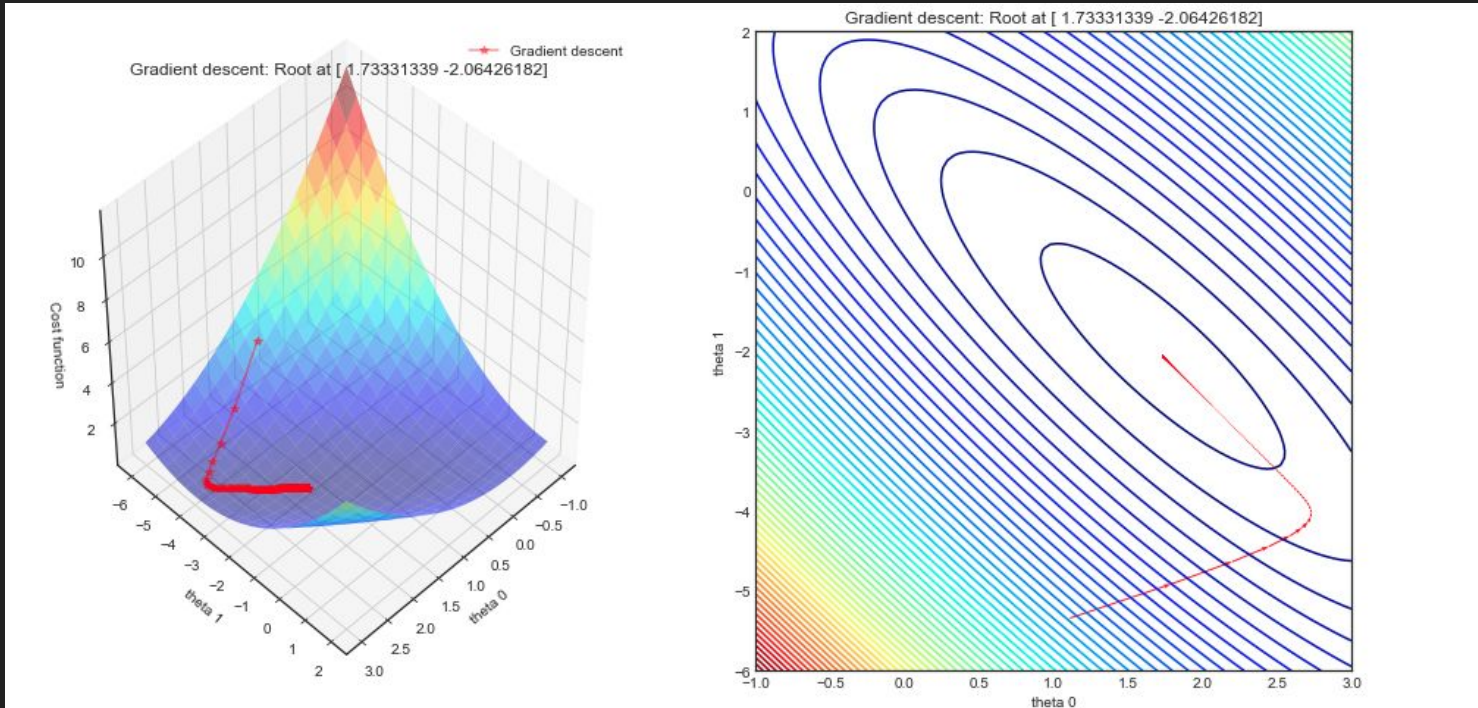


# Gradient Descent Algorithm

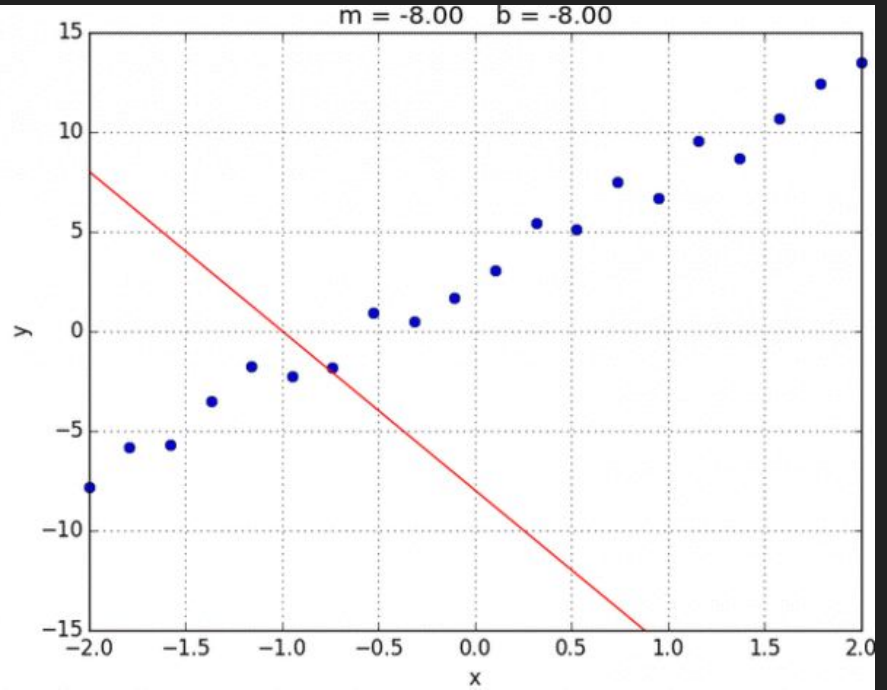
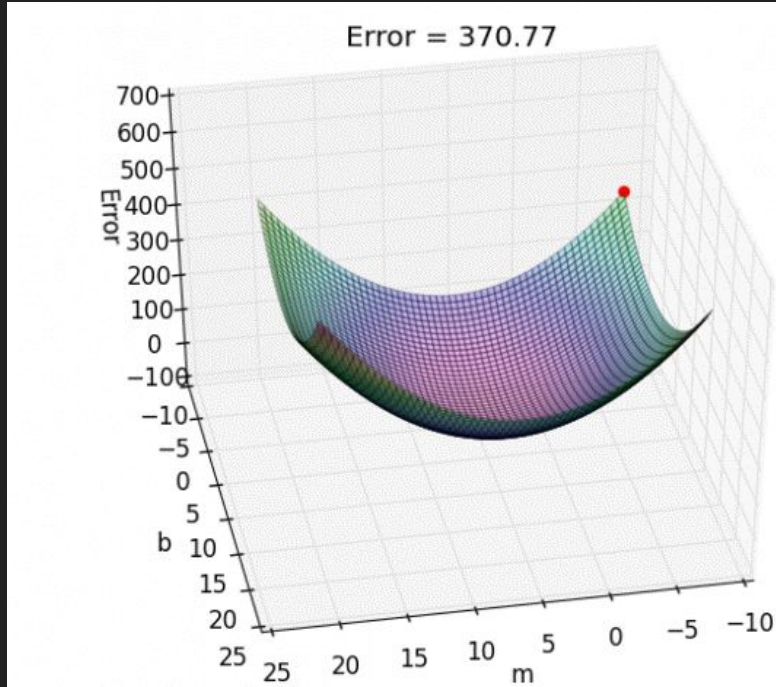




# Gradient Descent Algorithm



# Gradient Descent Algorithm



Q & A

# Gradient Descent Algorithm Hands On

- Visualization
  - <https://bl.ocks.org/EmilienDupont/aaf429be5705b219aaaf8d691e27ca87>
- Example
  - <https://github.com/arjun-kava/linear-regression-with-gradient-descent>
- Notebook
  - [https://colab.research.google.com/drive/1Dxs5fu3ELxF3B4P\\_Our6hO-pi7UQbLOH?usp=sharing](https://colab.research.google.com/drive/1Dxs5fu3ELxF3B4P_Our6hO-pi7UQbLOH?usp=sharing)