Data Science

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Recap

Unit 4

Machine Learning Algorithms

Objectives

- Supervised ML
- Unsupervised ML

Supervised Machine Learning

Machines are trained using well "labelled" training data and on basis of that data, machines predict the output.

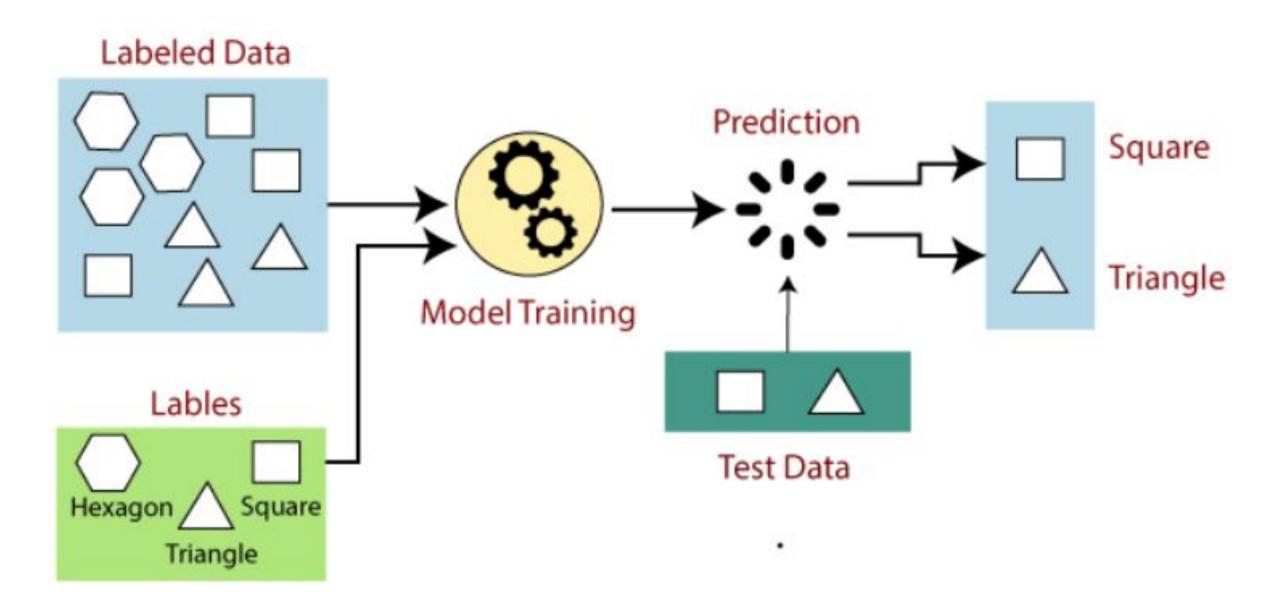
The labelled data means some input data is already tagged with the correct output.

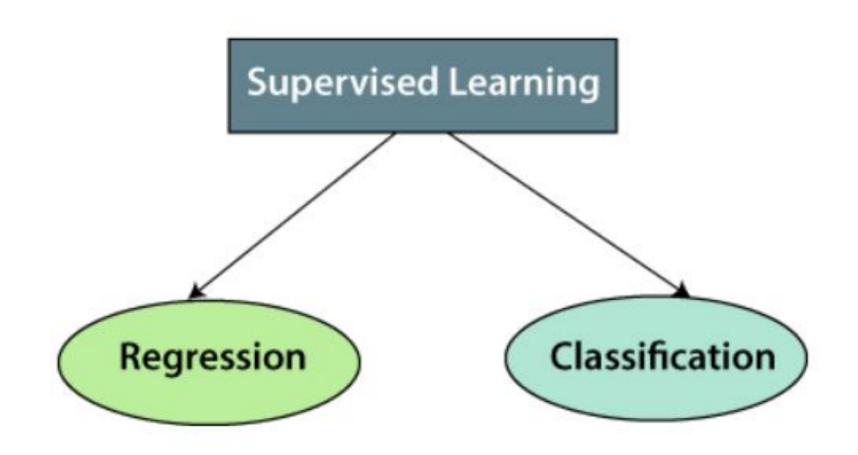
The training data provided to the machines work as the supervisor that teaches the machines to predict the output correctly

Supervised learning is a process of providing input data as well as correct output data to the machine learning model. The aim of a supervised learning algorithm is to find a mapping function to map the input variable(x) with the output variable(y).

Supervised learning can be used for Risk Assessment, Image classification, Fraud Detection, spam filtering, etc.

How Supervised Learning Works?





Some supervised learning algorithms:

- Linear Regression
- Regression Trees
- Non-Linear Regression
- Bayesian Linear Regression
- Polynomial Regression
- Random Forest
- Decision Trees
- Logistic Regression
- Support vector Machines

Unsupervised Machine Learning

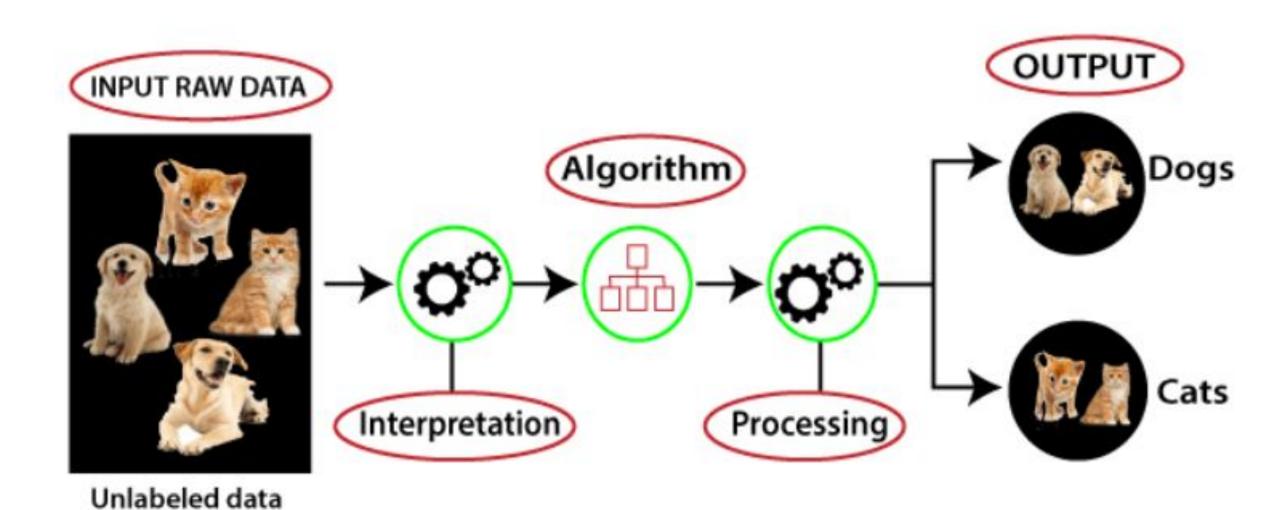
To find the hidden patterns from the given dataset. Models are not supervised using training dataset.

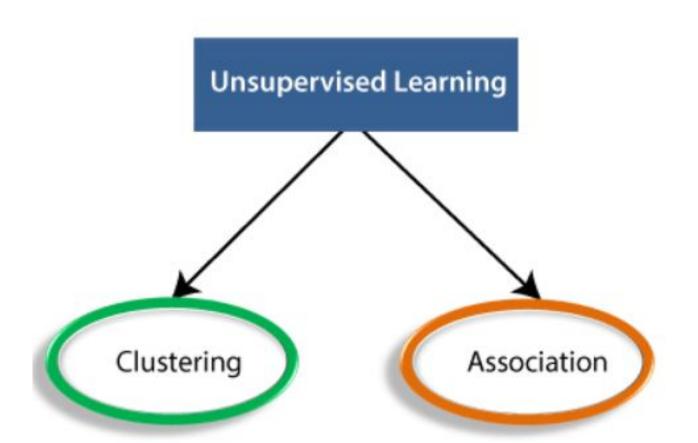
Instead, models itself find the hidden patterns and insights from the given data

Unsupervised learning is a type of machine learning in which models are trained using unlabeled dataset and are allowed to act on that data without any supervision.

Goal is to find the underlying structure of dataset, group that data according to similarities, and represent that dataset in a compressed format.

How Unsupervised Learning Works?





Some Unsupervised learning algorithms

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

Naïve Bayes Classifier Algorithm

Based on Bayes theorem and used for solving classification problems

It is mainly used in text classification that includes a high-dimensional training dataset.

It is a probabilistic classifier, which means it predicts on the basis of the probability of an object.

Some popular examples of Naïve Bayes Algorithm are spam filtration, Sentimental analysis, and classifying articles.

Why is it called Naïve Bayes?

Naïve: It is called Naïve because it assumes that the occurrence of a certain feature is independent of the occurrence of other features. Such as if the fruit is identified on the bases of color, shape, and taste, then red, spherical, and sweet fruit is recognized as an apple. Hence each feature individually contributes to identify that it is an apple without depending on each other.

Bayes: It is called Bayes because it depends on the principle of Bayes' Theorem.

Bayes' Theorem

Bayes' theorem is also known as Bayes' Rule or Bayes' law, which is used to determine the probability of a hypothesis with prior knowledge. It depends on the conditional probability.

The formula for Bayes' theorem is given as:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

Bayes' Theorem

 $P(A|B) = \frac{P(B|A)P(A)}{P(B)}$

Where,

P(A|B) is Posterior probability: Probability of hypothesis A on the observed event B.

P(B|A) is Likelihood probability: Probability of the evidence given that the probability of a hypothesis is true.

P(A) is Prior Probability: Probability of hypothesis before observing the evidence.

P(B) is Marginal Probability: Probability of Evidence.

Working Bayes' Theorem

Suppose we have a dataset of weather conditions and corresponding target variable "Play"

Need to decide that whether we should play or not on a particular day according to the weather conditions.

Problem: If the weather is sunny, then the Player should play or not? So to solve this problem, we need to follow the below steps:

- 1. Convert the given dataset into frequency tables.
- 2. Generate Likelihood table by finding the probabilities of given features.
- 3. Now, use Bayes theorem to calculate the posterior probability.

Frequency table for the Weather Conditions:

Weather	Yes	No
Overcast	5	0
Rainy	2	2
Sunny	3	2
Total	10	5

Likelihood table weather condition:

Weather	No	Yes	
Overcast	0	5	5/14= 0.35
Rainy	2	2	4/14=0.29
Sunny	2	3	5/14=0.35
All	4/14=0.29	10/14=0.71	

Applying Bayes'theorem:

P(Yes|Sunny)= P(Sunny|Yes)*P(Yes)/P(Sunny)P(Sunny|Yes)= 3/10= 0.3 P(Sunny)= 0.35 P(Yes)=0.71

So P(Yes | Sunny) = 0.3*0.71/0.35 =**0.60**

P(No|Sunny)= P(Sunny|No)*P(No)/P(Sunny)

P(Sunny|NO)= 2/4=0.5

P(No) = 0.29

P(Sunny)= 0.35

So P(No|Sunny) = 0.5*0.29/0.35 =**0.41**

So as we can see from the above calculation that P(Yes|Sunny)>P(No|Sunny) Hence on a Sunny day, Player can play the game.

Thank you