

## Machine Learning:

Machine learning, a branch of artificial intelligence, is a scientific discipline concerned with the design and development of algorithms that recognize complex patterns and make intelligent decisions based on input data.



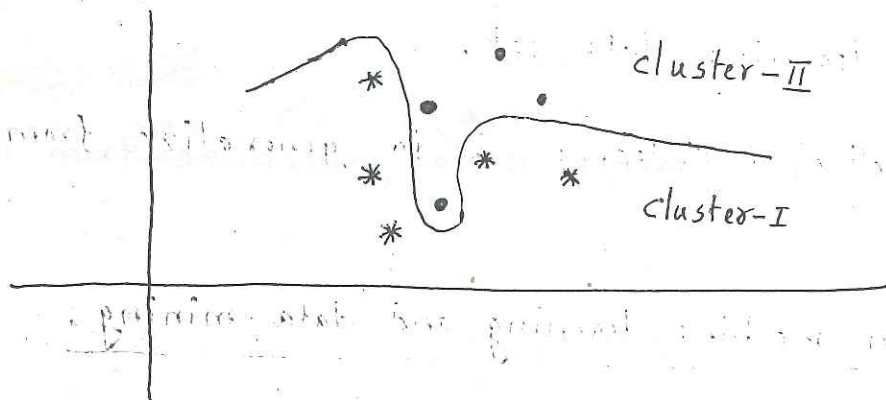
- Unsupervised learning
- Supervised learning

## Unsupervised learning:

In machine learning, unsupervised learning refers to the problem of trying to find hidden structure in unlabeled data.

### characteristics

- Input data.
- No labeled data or teacher signal.



### Approaches to unsupervised learning:

- K-mean clustering
- Modified K-mean clustering
- Hierarchical clustering
- Fuzzy C-Mean clustering

## Supervised Learning:

Supervised learning is the machine learning task of inferring a function from labeled training data.

Training data:

- Input object (a vector)
- Desired output value (supervisory signal)

A supervised learning algorithm analyzes the training data and produces an inferred function.

Inferred function is called

- classifier (if output is discrete)
- Regression function (if output is continuous)

Inferred function should predict the correct output value for any valid input object.

## Generalization:

- Generalization in this context is the ability of an algorithm to perform accurately on new, unseen examples after having trained on a learning data set.
- The core objective of a learner is to generalize from its experience.

## Difference between machine learning and data mining:

Machine learning:

Focuses on prediction based on known properties learned from the training data.

Data mining:

Discovery of unknown properties on the data.

## Approaches to Supervised learning:

- Artificial neural network.
- Bayesian network.
- Support Vector machines.
- Decision tree learning.
- Hidden Markov Model.
- Association rule learning.

## Unsupervised learning approaches:

### K-Mean clustering:

Classify the following data into two clusters using the K-mean algorithm.

$(3, 4)$   $(6, 7)$   $(4, 5)$   $(5, 7)$   $(2, 6)$

Ans:

(step 1) Initial step: (Centroid Selection)

class-I:  $(3, 4)$

class-II:  $(4, 5)$

step 2, 3, 4:

~~step 1~~  
Iteration 1: Data

	class-I $(3, 4)$	class-II $(4, 5)$	class
$(3, 4)$	0	<del>2.236</del> 1.414	class-I
$(6, 7)$	4.242	2.828	class-II
$(4, 5)$	<del>2.236</del> 1.414	0	class-II
$(5, 7)$	3.605	2.236	class-II
$(2, 6)$	2.236	2.236	class-I

$$\text{class-I: } (3, 4) (2, 6) \equiv \left( \frac{3+2}{2}, \frac{4+6}{2} \right) \\ \equiv (2.5, 5)$$

$$\text{class-II: } (6, 7) (4, 5) (5, 7) \equiv \left( \frac{6+4+5}{3}, \frac{7+5+7}{3} \right) \\ \equiv (5, 6.33)$$

Iteration 2:

~~steps~~

Data	class-I (2.5, 5)	class-II (5, 6.33)	class
(3, 4)	1.12	<del>3.07</del> 3.07	class-I
(6, 7)	4.03	1.208	class-II
(4, 5)	<del>1.5</del> 1.5	1.66	class-I
(5, 7)	3.2	0.67	class-II
(2, 6)	1.12	3.01	class-I

$$\text{class-I: } (3, 4) (4, 5) (2, 6) \equiv \left( \frac{3+4+2}{3}, \frac{4+5+6}{3} \right) \\ \equiv (3, 5)$$

$$\text{class-II: } (6, 7) (5, 7) \equiv \left( \frac{6+5}{2}, \frac{7+7}{2} \right) \\ \equiv (5.5, 7)$$



Iteration 3 :

Data	class-I (3, 5)	class-II (5, 5, 7)	class
(3, 4)	1.0	3.91	class-I
(6, 7)	3.61	0.50	class-II
(4, 5)	1.0	2.5	class-I
(5, 7)	2.83	<del>0.25</del> 0.5	class-II
(2, 6)	1.12	3.64	class-I

$$\text{class-I : } (3, 4) (4, 5) (2, 6) \equiv \left( \frac{3+4+2}{3}, \frac{4+5+6}{3} \right) \\ \equiv (3, 5)$$

$$\text{class-II : } (6, 7) (5, 7) \equiv \left( \frac{6+5}{2}, \frac{7+7}{2} \right) \\ \equiv (5.5, 7)$$

Algorithm stop

Output:

Data	class
(3, 4) (4, 5) (2, 6)	I
(6, 7) (5, 7)	II

## K-Mean clustering algorithm:

step1: Place  $K$  points the space represented by the objects that are being clustered. These points represent initial group centroids.

step2: Assign each object to the group that has the closest centroid.

step3: When all objects have been assigned, recalculate the positions of the  $K$  centroids.

step4: Repeat Step 2 and 3 until the centroids no longer move.

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### Modified K-Means clustering algorithm:

step1: Select initial group centroids - cleverly.

step2: Same as K-mean

step3: Same as K-mean

step4: Same as K-mean

Example using the previous data:

$$\frac{0.4325}{4.73}$$

step 1: Initial centroid selection

Data	(3, 4)	(6, 7)	(4, 5)	(5, 7)	(2, 6)
(3, 4)	0	4.242	1.414	3.605	2.236
(6, 7)		0	2.828	1	4.123
(4, 5)			0	2.24	2.24
(5, 7)				0	3.162
(2, 6)					0

class-I: (3, 4)

class-II: (6, 7)

step 2, 3, 4:

Iteration-I:

Data	class-I (3, 4)	class-II (6, 7)	class
(3, 4)	0	4.242	class-I
(6, 7)	4.242	0	class-II
(4, 5)	1.414	2.828	class-I
(5, 7)	3.605	1	class-II
(2, 6)	2.236	4.123	class-I



$$\text{class-I: } (3, 4) (4, 5) (2, 6) \equiv \left( \frac{3+4+2}{3}, \frac{4+5+6}{3} \right) \\ \equiv (3, 5)$$

$$\text{class-II: } (6, 7) (5, 7) \equiv \cancel{(6, 7)} \left( \frac{6+5}{2}, \frac{7+7}{2} \right) \\ \equiv (5.5, 7)$$

Iteration-II:

Data	class-I (3, 5)	class-II (5.5, 7)	class
(3, 4)	1	3.91	class-I
(6, 7)	3.61	0.5	class-II
(4, 5)	1	<del>6.75</del> 2.5	class-I
(5, 7)	2.83	0.5	class-II
(2, 6)	1.414	3.64	class-I

$$\text{class-I: } (3, 4) (4, 5) (2, 6) \equiv \left( \frac{3+4+2}{3}, \frac{4+5+6}{3} \right) \\ \equiv (3, 5)$$

$$\text{class-II: } (6, 7) (5, 7) \equiv \left( \frac{6+5}{2}, \frac{7+7}{2} \right) \\ \equiv (5.5, 7)$$

Algorithm stop

Data	class	centroid
(3, 4) (4, 5) (2, 6)	class-I	(3, 5)
(6, 7) (5, 7)	class-II	(5.5, 7)