

ASSIGNMENT – 2 MACHINE LEARNING

Q1 to Q11 have only one correct answer. Choose the correct option to answer your question.

1. Movie Recommendation systems are an example of:

- i) Classification
- ii) Clustering
- iii) Regression

ANS: A

- a) 2 Only
- b) 1 and 2
- c) 1 and 3
- d) 2 and 3

2. Sentiment Analysis is an example of:

- i) Regression
- ii) Classification
- iii) Clustering
- iv) Reinforcement

ANS: D

- a) 1 Only
- b) 1 and 2
- c) 1 and 3
- d) 1, 2 and 4

3. Can decision trees be used for performing clustering?

- a) True
- b) False

ANS: A

4. Which of the following is the most appropriate strategy for data cleaning before performing clustering analysis, given less than desirable number of data points:

- i) Capping and flooring of variables

ii) Removal of outliers

ANS: A

a) 1 only

b) 2 only

c) 1 and 2

d) None of the above

5. What is the minimum no. of variables/ features required to perform clustering?

a) 0

b) 1

c) 2

d) 3

ANS: B

6. For two runs of K-Mean clustering is it expected to get same clustering results?

a) Yes

b) No

ANS: B

7. Is it possible that Assignment of observations to clusters does not change between successive iterations in K-Means?

a) Yes

b) No

c) Can't say

d) None of these

ANS: A

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8. Which of the following can act as possible termination conditions in K-Means?

i) For a fixed number of iterations.

ii) Assignment of observations to clusters does not change between iterations. Except for cases with a bad local minimum.

iii) Centroids do not change between successive iterations.

iv) Terminate when RSS falls below a threshold.

ANS: D

a) 1, 3 and 4

b) 1, 2 and 3

c) 1, 2 and 4

d) All of the above

9. Which of the following algorithms is most sensitive to outliers?

a) K-means clustering algorithm

b) K-medians clustering algorithm

c) K-modes clustering algorithm

d) K-medoids clustering algorithm

ANS: A

10. How can Clustering (Unsupervised Learning) be used to improve the accuracy of Linear Regression model (Supervised Learning):

i) Creating different models for different cluster groups.

ii) Creating an input feature for cluster ids as an ordinal variable.

iii) Creating an input feature for cluster centroids as a continuous variable.

iv) Creating an input feature for cluster size as a continuous variable.

Options:

a) 1 only

b) 2 only

c) 3 and 4

d) All of the above

ANS: D

11. What could be the possible reason(s) for producing two different dendrograms using agglomerative clustering algorithms for the same dataset?

- a) Proximity function used
- b) of data points used
- c) of variables used
- d) All of the above

ANS: D

Q12 to Q14 are subjective answers type questions, Answers them in their own words briefly

12. Is K sensitive to outliers?

ANS: K is an unsupervised learning algorithm which aims to partition n observations into k clusters in which each observation belongs to the cluster with the nearest centroid. The algorithm aims to minimize the squared Euclidean distances between the observation and the centroid of cluster to which it belongs.

But sometime K-Means algorithm does not give best results. It is sensitive to outliers. An outlier is a point which is different from the rest of data points

13. Why is K means better?

Ans: k-means is the fastest and most efficient algorithm to categorize data points into groups even when very little information is available about data .also its having some advantages that are given bellow

K means is very simple to implement.

Generalization of clusters for different shape and size.

It adapts the new examples very frequently.

K means is scalable to a huge data set and also faster to large datasets.

14. Is K means a deterministic algorithm?

ANS: The basic k-means clustering is based on a **non-deterministic algorithm**. This means that running the algorithm several times on the same data, could give different results.

The non-deterministic nature of K-Means is due to its random selection of data points as initial centroids. The key idea of the algorithm is to select data points which belong to dense regions and which are adequately separated in feature space as the initial centroids.