- Q 1. Movie Recommendation systems are an example of:
- i) Classification
- ii) Clustering
- iii) Regression

Ans - b) 1 and 2

Q 2. Sentiment Analysis is an example of:

Ans - d) 1, 2 and 4

Q 3. Can decision trees be used for performing clustering?

Ans - a) True

- Q 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 Options:

Ans - a) 1 only

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

Ans - b) 1

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

Ans - b) No

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

Ans - a) Yes

- Q 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 witha bad local minimum.
- iii) Centroids do not change between successive iterations.
- iv) Terminate when RSS falls below a threshold.

Ans - d) All of the above

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

Ans - a) K-means clustering algorithm

- Q 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.

Ans - d) All of the above

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

Ans - d) All of the above

Q 12. Is K sensitive to outliers?

Ans - *k*-Means is a well studied clustering problem that finds applications in many fields related to <u>unsupervised learning</u>. *k*-means clustering is "highly sensitive" to the isolated points. These isolated points called outliers.

For example -

is the average of the salaries of the following people:

50k, 20k, 35k, 65k and 1MM

The average ends up being (50k + 20k + 35k + 65k + 1MM) / 5 = 1170k / 5 = 234k.

If we did not have the 1MM outlier, the average would have been 10k + 20k + 35k + 65k / 4 = 170k / 4 = 42.5k.

Note that the two average results are wildly different from one another.

So we can say k-mean algorithm is not making sense to use.

Also we can say k-means clustering is Highly sensitive to outliers.

Q 13. Why is K means better?

Ans -

- 1) Relatively simple to implement.
- 2) Scales to large data sets.
- 3) Guarantees convergence.
- 4) Can warm-start the positions of centroids.
- 5) Easily adapts to new examples.
- 6) Generalizes to clusters of different shapes and sizes, such as elliptical clusters.
- Q 14. Is K means a deterministic algorithm?

 ${\rm Ans-No}$, Nature of K-Means is due to its random selection of data points as initial centroids making "non- deterministic algorithm ".

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. However, to ensure consistent results.