

## Dimensionality Reduction MCQ (Version : 1)

TEST

● **Correct Answer**

🕒 Answered in 238.23333333333 Minutes

### Question 1/17

Principal Component Analysis is a linear dimensionality reduction technique.

☐ False

☒ True

### Question 2/17

Before performing PCA, what should ideally be done to the data set?

☒ standardize variables

☐ categorise variables

☐ encode variables

☐ normalise variables

### Question 3/17

What can be said of the largest eigenvalue, in terms of its relation to the principal components of a dataset?

☐ it is equal to the amount of variance in the

☐ last principal component

☐ it is equal to the amount of variance in the first principal component

☐ it corresponds to the direction in which there is the least amount of variance in the data

☒ it corresponds to the direction in which there is the largest amount of variance in the data

### Question 4/17

Which of the following is NOT a method for dimensionality reduction?

☐ t-SNE

☐ Principal Component Analysis

☐ Multidimensional Scaling

☒ k-means clustering

### Question 5/17

t-SNE is a linear dimensionality reduction technique.

☒ False

☐ True

## Question 6/17

Which of the following is not true regarding dimensionality reduction:

☐ it decreases the interpretability of models

☐ it can be used to aid data visualization

☒ it always improves performance of clustering algorithms

☐ it decreases the computational time for training models

## Question 7/17

t-SNE has one tunable hyperparameter, namely:

☒ perplexity

☐ duplicity

☐ irregularity

☐ complexity

## Question 8/17

Which of the following dimensionality reduction techniques preserves distances between points?

☒ Multidimensional scaling

☐ t-SNE

## Question 9/17

It is not necessary to have a target variable for applying dimensionality reduction algorithms.

☒ True

☐ False

## Question 10/17

The most popularly used dimensionality reduction algorithm is Principal Component Analysis (PCA). Which of the following is/are true about PCA?

1. PCA is an unsupervised method
2. It searches for the directions that data have the largest variance
3. Maximum number of principal components  $\leq$  number of features
4. All principal components are orthogonal to each other

☒ 1, 2, 3 and 4

☐ 1, 2, and 4

☐ 1, 2 and 3

☐ 1, 3 and 4

## Question 11/17

In which of the following scenarios is t-SNE better to use than PCA for dimensionality reduction while working on a local machine with minimal computational power?

☐ Data set with 10,000 entries and 200 features

☐ Data set with 1 Million entries and 300 features



Data set with 10,000 entries and 8 features



Data set with 100000 entries and 310 features

## Question 12/17

In t-SNE algorithm, which of the following hyper parameters can be tuned?



Number of dimensions



Smooth measure of effective number of neighbours



Number of dimensions, smooth measure and max iterations



Maximum number of iterations

## Question 13/17

Which of the following statement is correct for t-SNE and PCA?



t-SNE is linear whereas PCA is non-linear



t-SNE and PCA both are linear



t-SNE and PCA both are nonlinear



t-SNE is nonlinear whereas PCA is linear

## Question 14/17

What will happen when eigenvalues are roughly equal?

☐

The number of dimensions in the data will increase

☐

PCA will perform optimally



PCA will perform poorly

☐

All eigenvectors will be the same

## Question 15/17

An embedding is a representation of a vector in a different feature space.



True

☐

False

## Question 16/17

In Multi-dimensional Scaling, a larger stress value is preferred, and it is this quantity that is maximised by MDS.

☐

True



False

## Question 17/17

Regarding an MDS scatter plot, which of the following is false:



The distances between observations are not preserved



The axes themselves are meaningless



The important thing is the proximity of the points



The orientation of the figure is arbitrary