Ans. 1 (C)

Ans. 2 (C)

Ans. 3 (A)

Ans. 4 (B)

Ans. 5 (B)

Ans. 6 (C)

Ans. 7 (C)

Ans. 8 (B,C)

Ans. 9 (A,D)

Ans. 10 (A,B,D)

Ans. 12

Bagging	Boosting
Various training data subsets are randomly drawn with replacement from the whole training dataset.	Each new subset contains the components that were misclassified by previous models.
Bagging attempts to tackle the over-fitting issue.	Boosting tries to reduce bias.
If the classifier is unstable (high variance), then we need to apply bagging.	If the classifier is steady and straightforward (high bias), then we need to apply boosting.
Every model receives an equal weight.	Models are weighted by their performance.
Objective to decrease variance, not bias.	Objective to decrease bias, not variance.
It is the easiest way of connecting predictions that belong to the same type.	It is a way of connecting predictions that belong to the different types.
Every model is constructed independently.	New models are affected by the performance of the previously developed model.

Ans. 13 The R-squared statistic provides a measure of fit. It takes the form of a proportion—the proportion of variance explained—and so it always takes on avalue between 0 and 1. In simple words, it represents how much of our data is being explained by our model. For example,

statistic = 0.75, itsays that our model fits 75 % of the total data set. Similarly, if it is 0, it means none of the data points is being explained and a value of 1 represents100% data explanation. Mathematically

$$R^2 = \frac{\text{TSS} - \text{RSS}}{\text{TSS}} = 1 - \frac{\text{RSS}}{\text{TSS}}$$

statistic is calculated as:

Ans. 14 In Normalisation, the change in values is that they are at a standard scale without distorting the differences in the values. Whereas, Standardisation assumes that the dataset is in Gaussian distribution and measures the variable at different scales, making all the variables equally contribute to the analysis.

Ans. 15 Cross-Validation is a statistical method of evaluating and comparing learning algorithms by dividing data into two segments: one used to learn or train a model and the other used to validate the model.

The purpose of cross—validation is to test the ability of a machine learning model to predict new data. It is also used to flag problems like overfitting or selection bias and gives insights on how the model will generalize to an independent dataset.

The disadvantage of this method is that the training algorithm has to be rerun from scratch k times, which means it takes k times as much computation to make an evaluation. A variant of this method is to randomly divide the data into a test and training set k different times.