

## DA -04

1. Which of the following are model parameters, not hyperparameters?

- ☐ (A) Weights in Logistic Regression
- ☐ (B) Regularization strength C in Logistic Regression
- ☐ (C) max\_depth in Decision Tree
- ☐ (D) Support vectors in SVM

2. In a Decision Tree classifier, which hyperparameter prevents overfitting by limiting tree complexity?

- ☐ (A) max\_depth
- ☐ (B) min\_samples\_split
- ☐ (C) min\_samples\_leaf
- ☐ (D) All of the above

3. When tuning SVM with RBF kernel, what is the effect of increasing gamma?

- ☐ (A) Makes decision boundary smoother (less overfitting)
- ☐ (B) Makes decision boundary more complex (risk of overfitting)
- ☐ (C) Increases margin width
- ☐ (D) Has no effect

4. Which hyperparameter in Logistic Regression controls the amount of regularization?

- ☐ (A) penalty
- ☐ (B) solver
- ☐ (C) C
- ☐ (D) max\_iter

- 5.** You run a GridSearchCV with a very large parameter grid. Which of the following is MOST likely?
- ☐ (A) Fast search but poor performance
  - ☐ (B) Very slow search but exhaustive
  - ☐ (C) Random exploration of parameter space
  - ☐ (D) Tuning may fail due to no refitting
- 6.** Which statement about GridSearchCV is FALSE?
- ☐ (A) It can use multiple scoring metrics
  - ☐ (B) It can perform cross-validation
  - ☐ (C) It automatically chooses hyperparameters during training updates
  - ☐ (D) It can refit the best model
- 7.** Which hyperparameter in SVM controls the trade-off between margin size and misclassification errors?
- ☐ (A) gamma
  - ☐ (B) kernel
  - ☐ (C) C
  - ☐ (D) degree
- 8.** You use GridSearchCV with cv=5 on a dataset of 1000 samples. How many models are trained per hyperparameter combination?
- ☐ (A) 5
  - ☐ (B) 1000
  - ☐ (C) 200
  - ☐ (D) 1
- 9.** Which advanced method discards poorly performing hyperparameter configurations early to save time?
- ☐ (A) Grid Search
  - ☐ (B) Random Search
  - ☐ (C) Hyperband / Successive Halving
  - ☐ (D) Cross-validation

- 10.** If you include the test set during hyperparameter tuning, what problem occurs?
- ☐ (A) Data Leakage
  - ☐ (B) Over-regularization
  - ☐ (C) Underfitting
  - ☐ (D) Faster convergence
- 11.** You are tuning Logistic Regression for the Breast Cancer Wisconsin dataset. You try values of  $C = [0.01, 0.1, 1, 10]$ . When  $C = 0.01$ , the model underfits (low train and test accuracy). What is the correct reason?
- ☐ (A)  $C=0.01$  applies strong regularization, shrinking coefficients too much.
  - ☐ (B)  $C=0.01$  applies weak regularization, allowing overfitting.
  - ☐ (C) Logistic regression cannot handle this dataset.
  - ☐ (D)  $C$  has no effect on model complexity.
- 12.** You train a Decision Tree without setting `max_depth`, and it achieves 100% training accuracy but only 70% test accuracy. What hyperparameter should you tune first to reduce overfitting?
- ☐ (A) `criterion`
  - ☐ (B) `max_depth`
  - ☐ (C) `splitter`
  - ☐ (D) `random_state`
- 13.** On the Breast Cancer dataset, you use an SVM with RBF kernel. With  $\gamma = 0.0001$ , the model performs poorly (both train and test accuracy are low). What is the most likely reason?
- ☐ (A)  $\gamma$  is too low, decision boundary is too simple.
  - ☐ (B)  $\gamma$  is too high, causing overfitting.
  - ☐ (C) Kernel is wrong; should use linear kernel.
  - ☐ (D)  $C$  value is too high.

- 14.** You perform GridSearchCV with the following parameter grid for Decision Tree: `param_grid = {"max_depth": [3, 5, 7, None], "min_samples_split": [2, 5, 10]}` If `cv=5` and your dataset has 1000 samples, how many models are trained in total?
- ☐ (A) 12
  - ☐ (B) 60
  - ☐ (C) 20
  - ☐ (D) 15
- 15.** While tuning hyperparameters for Logistic Regression, a student mistakenly used the test set inside GridSearchCV. Which of the following is MOST likely to happen?
- ☐ (A) Test accuracy will be underestimated.
  - ☐ (B) Model will overfit training data only.
  - ☐ (C) Hyperparameters will be biased toward test set, giving overly optimistic results.
  - ☐ (D) Nothing significant, since CV already prevents overfitting.
- 16.** For an SVM with RBF kernel, you want to tune C and gamma. The search space is very large. Which tuning method is most efficient to try first?
- ☐ (A) Grid Search
  - ☐ (B) Random Search
  - ☐ (C) Manual search
  - ☐ (D) Successive Halving
- 17.** You run GridSearchCV with Logistic Regression and find that the best `C=1.0`. How do you get the best model to use for predictions?
- ☐ (A) `grid.best_params_`
  - ☐ (B) `grid.best_score_`
  - ☐ (C) `grid.best_estimator_`
  - ☐ (D) `grid.cv_results_`
- 18.** You restrict `max_depth=2` for a Decision Tree on Breast Cancer dataset, but accuracy is only ~65% on both train and test sets. What is happening?
- ☐ (A) Overfitting
  - ☐ (B) Underfitting
  - ☐ (C) Data Leakage
  - ☐ (D) High variance

**19.** Which hyperparameter in SVM directly controls how much misclassification is tolerated?

- ☐ (A) gamma
- ☐ (B) degree
- ☐ (C) kernel
- ☐ (D) C

**20.** Which of the following best describes ensemble learning?

- ☐ (A) Training a single model with more epochs to improve accuracy
- ☐ (B) Combining multiple models to achieve better performance
- ☐ (C) Using deep learning models instead of shallow ones
- ☐ (D) Training models only in parallel on different machines

**21.** What is the main idea behind bagging?

- ☐ (A) Assigning higher weights to misclassified examples
- ☐ (B) Combining weak learners sequentially
- ☐ (C) Training base learners on different random subsets of data with replacement
- ☐ (D) Reducing bias by averaging predictions from one model

**22.** In boosting, each new learner is trained to:

- ☐ (A) Reduce correlation between models
- ☐ (B) Correct mistakes made by previous learners
- ☐ (C) Train on a completely random subset of data
- ☐ (D) Maximize variance of predictions

**23.** Which of the following ensemble methods is an example of parallel ensemble learning?

- ☐ (A) AdaBoost
- ☐ (B) Gradient Boosting
- ☐ (C) Random Forest
- ☐ (D) XGBoost

**24.** What type of sampling is used in bagging?

- ☐ (A) Random sampling without replacement
- ☐ (B) Random sampling with replacement (bootstrap sampling)
- ☐ (C) Stratified sampling
- ☐ (D) Systematic sampling

- 25.** In Random Forest, why is feature randomness (`max_features`) used?
- ☐ (A) To reduce the training time of each tree
  - ☐ (B) To ensure each tree is different and less correlated
  - ☐ (C) To increase the number of leaf nodes
  - ☐ (D) To decrease the variance of each tree individually
- 26.** Which of the following is a major disadvantage of ensemble learning?
- ☐ (A) It always increases bias
  - ☐ (B) It is less interpretable and computationally more expensive
  - ☐ (C) It cannot handle non-linear data
  - ☐ (D) It does not reduce variance
- 27.** In boosting, if a data point is misclassified repeatedly, what usually happens?
- ☐ (A) Its weight decreases in the next iteration
  - ☐ (B) Its weight increases in the next iteration
  - ☐ (C) It gets removed from the dataset
  - ☐ (D) The model ignores it
- 28.** Which of the following pairs correctly matches method with base idea?
- ☐ (A) Bagging – Reduce bias
  - ☐ (B) Boosting – Reduce variance
  - ☐ (C) Bagging – Reduce variance
  - ☐ (D) Boosting – Reduce interpretability only
- 29.** You are training a model for a medical diagnosis dataset (high-stakes, small dataset). The base decision tree tends to overfit. Which ensemble method would be more suitable?
- ☐ (A) Bagging
  - ☐ (B) Boosting
  - ☐ (C) Logistic Regression
  - ☐ (D) K-Means

**30.** In clustering, the goal is to

- ☐ (A) Minimize within-cluster similarity
- ☐ (B) Maximize between-cluster similarity
- ☐ (C) Both (a) and (b)
- ☐ (D) None of the above

**31.** K-Means clustering requires

- ☐ (A) Pre-defined number of clusters (k)
- ☐ (B) Distance metric
- ☐ (C) Random initialization of centroids
- ☐ (D) All of the above

**32.** The main drawback of K-Means clustering is

- ☐ (A) It can only handle categorical data
- ☐ (B) It requires knowing the number of clusters in advance
- ☐ (C) It is not suitable for large datasets
- ☐ (D) It doesn't use distance metrics

**33.** Which metric is commonly used to evaluate clustering results without ground truth labels

- ☐ (A) Accuracy
- ☐ (B) Adjusted Rand Index
- ☐ (C) Silhouette Score
- ☐ (D) F1 Score

**34.** Dimensionality reduction is mainly used to

- ☐ (A) Increase the number of features
- ☐ (B) Reduce noise and redundancy in data
- ☐ (C) Increase model complexity
- ☐ (D) Improve overfitting

**35.** In PCA, the new axes (principal components) are

- ☐ (A) Original features scaled
- ☐ (B) Orthogonal linear combinations of features
- ☐ (C) Random projections of features
- ☐ (D) Nonlinear transformations

**36.** The first principal component (PC1) captures

- ☐ (A) The smallest variance in the data
- ☐ (B) The largest variance in the data
- ☐ (C) Only categorical variance
- ☐ (D) No variance at all

**37.** Eigenvalues in PCA represent

- ☐ (A) Variance explained by each principal component
- ☐ (B) The angle between components
- ☐ (C) The mean of the dataset
- ☐ (D) The correlation coefficient

**38.** Which step comes first in PCA

- ☐ (A) Compute eigenvalues/eigenvectors
- ☐ (B) Standardize the dataset
- ☐ (C) Project data onto new space
- ☐ (D) Compute covariance matrix

**39.** If two variables are highly correlated, PCA will

- ☐ (A) Drop one variable
- ☐ (B) Merge them into a principal component
- ☐ (C) Ignore both variables
- ☐ (D) Always assign them to PC2

**40.** The curse of dimensionality generally refers to

- ☐ (A) Increase in computation cost as data grows in rows
- ☐ (B) Problems when data has too many features
- ☐ (C) Errors in model labeling
- ☐ (D) Lack of training samples