

Topics: Introduction • ANN • ARM • Clustering • Trees • Regression • Assessment • SVM • Preprocessing • Probabilistic • Advanced

Which statement best describes the “generalization gap” of a model?

- A. Difference between training loss and validation/test loss
- B. Difference between precision and recall
- C. Difference between accuracy and F1-score
- D. Difference between feature scaling methods

A model has high training error and high validation error. What is the most likely issue?

- A. Underfitting (high bias)
- B. Overfitting (high variance)
- C. Data leakage from test to train
- D. Too much regularization is impossible

Which of the following is a hyperparameter (not learned directly from training loss minimization)?

- A. Learning rate in gradient descent
- B. Weight vector w in linear regression
- C. Bias term b in logistic regression
- D. Predicted label \hat{y} for each sample

Which choice is the best reason to keep a separate “test set”?

- A. To estimate final performance on unseen data after model selection
- B. To tune hyperparameters (e.g., λ , C)
- C. To make training faster
- D. To increase the number of training samples

Which tasks are typically unsupervised? (Choose ALL that apply)

- A. Customer segmentation based on purchase vectors
- B. Predicting house price from square footage
- C. Discovering topics in a collection of documents
- D. Classifying emails as spam or not spam

Why are all weights in a multi-layer neural network NOT initialized to zero?

- A. Zero initialization keeps neurons symmetric so they learn the same features
- B. Zero initialization makes the loss function non-differentiable
- C. Zero initialization prevents using mini-batch training
- D. Zero initialization forces gradients to explode

In backpropagation, which quantity is propagated from later layers to earlier layers?

- A. Gradient of the loss with respect to activations (error signal)
- B. The raw input features x
- C. The learning rate η
- D. The one-hot labels y

Which activation function outputs values strictly in the range $(0, 1)$?

- A. Sigmoid
- B. ReLU
- C. Tanh
- D. Identity (linear)

Which technique reduces overfitting by randomly “dropping” units during training?

- A. Dropout
- B. Batch normalization
- C. Momentum
- D. Early stopping always increases overfitting

If we increase the minimum confidence (minconf) while keeping minsup fixed, what usually happens?

- A. Fewer association rules pass the threshold
- B. More frequent itemsets are generated
- C. Support of all itemsets increases
- D. Apriori needs fewer database scans regardless of max itemset size

Which item is NOT required to run the Apriori algorithm?

- A. Class labels for each transaction
- B. A transaction database
- C. A minimum support threshold
- D. A method to count itemset supports

Suppose $\text{support}(X)=0.20$ and $\text{support}(X \cup Y)=0.05$. What is $\text{confidence}(X \rightarrow Y)$?

- A. 0.25
- B. 0.05
- C. 0.15
- D. 4.00

Lift($X \rightarrow Y$) is defined as:

- A. $\text{confidence}(X \rightarrow Y) / \text{support}(Y)$
- B. $\text{support}(X \cup Y) + \text{support}(X)$
- C. $\text{support}(X) / \text{support}(Y)$
- D. $\text{confidence}(X \rightarrow Y) - \text{support}(Y)$

Apriori anti-monotone property implies: if an itemset is infrequent, then all its supersets are:

- A. Infrequent
- B. Frequent
- C. Candidates in the next iteration
- D. Guaranteed to have higher confidence

A dataset has 10 transactions. $\text{support}(\{A,B\})=4/10$, $\text{support}(\{C\})=3/10$, and $\text{support}(\{A,B,C\})=3/10$. With $\text{minsup}=30\%$ and $\text{minconf}=80\%$, which rule is valid?

- A. $\{A,B\} \rightarrow \{C\}$
- B. $\{C\} \rightarrow \{A,B\}$
- C. $\{A\} \rightarrow \{B,C\}$
- D. No rule is valid

K-means clustering chooses centroids to minimize:

- A. Sum of squared distances from points to their assigned centroid
- B. Number of misclassified labels
- C. Total absolute correlation between features
- D. Entropy of the class distribution

Before running k-means with Euclidean distance on features with very different scales, you should usually:

- A. Standardize/normalize features
- B. Convert labels to one-hot vectors
- C. Increase K until training loss is zero
- D. Remove all outliers so every cluster is spherical

1D points: {1, 2, 9, 10}. Initial centroids: $c1=1$, $c2=10$. After one k-means update (assign then recompute), what are the new centroids?

- A. $c1=1.5$, $c2=9.5$
- B. $c1=2.0$, $c2=10.0$
- C. $c1=3.0$, $c2=9.0$
- D. $c1=1.0$, $c2=10.0$

Which statements about k-means are TRUE? (Choose ALL that apply)

- A. The objective value (WCSS) never increases after an assignment+update iteration
- B. K-means always finds the global optimum regardless of initialization
- C. Increasing K can only decrease (or keep) the optimal WCSS on the training data
- D. K-means naturally handles non-convex clusters (e.g., two-moons) without issues

Information gain for a split is based on the reduction of:

- A. Impurity (e.g., entropy or Gini) from parent to children
- B. Learning rate from epoch to epoch
- C. Number of features used in the model
- D. Regularization strength in the loss function

In a random forest, why is a random subset of features often considered at each split?

- A. To reduce correlation between trees and improve ensemble variance reduction
- B. To guarantee perfect accuracy on training data
- C. To make each tree deterministic
- D. To ensure all trees use the same splits

Which approach is commonly used to reduce overfitting in a decision tree?

- A. Pruning (pre-pruning or post-pruning)
- B. Using a larger max depth with no stopping criteria
- C. Removing the training set and training on the test set
- D. Always choosing splits with the smallest information gain

Gain ratio is often preferred over information gain because it:

- A. Reduces bias toward attributes with many distinct values
- B. Eliminates the need to compute entropy
- C. Works only for regression trees
- D. Always produces shallower trees

Compared to k-NN and SVM, a key advantage of decision trees is that they:

- A. Are easy to interpret as a set of if-then rules
- B. Always outperform ensembles on test data
- C. Require feature scaling to work correctly
- D. Cannot handle categorical features

A dataset has entropy $H(\text{parent})=1.0$. Split by feature A gives weighted child entropy 0.60. Split by feature B gives weighted child entropy 0.72. Which split has higher information gain?

- A. Feature A
- B. Feature B
- C. Both equal
- D. Cannot be determined without knowing the number of samples

Which regularization is most likely to produce sparse coefficients (some exactly zero)?

- A. L1 (Lasso) regularization
- B. L2 (Ridge) regularization
- C. No regularization
- D. Early stopping guarantees sparsity

In linear regression, the coefficient of determination R^2 is commonly interpreted as:

- A. The fraction of variance in y explained by the model
- B. The slope of the best-fit line
- C. The average absolute error
- D. The probability the model is correct

Ordinary Least Squares (OLS) fits parameters by minimizing:

- A. Sum of squared residuals
- B. Sum of absolute residuals
- C. Classification error rate
- D. Hinge loss

A key difference between linear regression and logistic regression is that logistic regression:

- A. Models $P(y=1 | x)$ using a sigmoid (or similar) link and is used for classification
- B. Always yields a closed-form solution
- C. Requires categorical features only
- D. Minimizes squared error on real-valued targets

Q30. Linear Regression / Regression

Hard • Multiple

For ridge regression, increasing the regularization strength λ typically: (Choose ALL that apply)

- A. Increases bias
- B. Decreases variance
- C. Makes coefficients larger in magnitude
- D. Moves coefficients toward zero

For an imbalanced classification dataset, which cross-validation variant is most appropriate?

- A. Stratified k-fold (preserve class proportions in each fold)
- B. Random k-fold without constraints
- C. Leave-one-out only
- D. Train-test split without shuffling

Precision for the positive class is defined as:

- A. $TP / (TP + FP)$
- B. $TP / (TP + FN)$
- C. $TN / (TN + FP)$
- D. $(TP + TN) / (TP + TN + FP + FN)$

In a disease screening task, false negatives are extremely costly (missing a sick patient). Which metric should you prioritize when selecting a threshold?

- A. Recall (sensitivity)
- B. Precision
- C. Specificity only
- D. Overall accuracy

In soft-margin SVM, the hyperparameter C mainly controls:

- A. Trade-off between margin width and classification errors (slack penalties)
- B. Number of hidden layers
- C. Number of clusters used in training
- D. Learning rate schedule

In a hard-margin linear SVM, which training points become support vectors?

- A. Points that lie on the margin boundaries
- B. All points in the majority class
- C. Only misclassified points
- D. All points farthest from the hyperplane

Why should you compute scaling parameters (mean/variance) using only the training set?

- A. To avoid data leakage from validation/test into training
- B. Because scaling on all data is mathematically incorrect
- C. Because models cannot handle scaled validation data
- D. Because scaling changes labels

A common simple strategy for missing values in a numeric feature is:

- A. Impute with the mean/median of the training feature
- B. Replace missing values with random class labels
- C. Drop all columns that contain any missing value
- D. Always set missing values to 0 (no exceptions)

Compared to Maximum Likelihood Estimation (MLE), Maximum A Posteriori (MAP) estimation:

- A. Incorporates a prior distribution over parameters
- B. Ignores observed data
- C. Always produces the same estimate as MLE
- D. Requires no assumptions about parameters

The “naive” assumption in Naive Bayes classifier is that features are:

- A. Conditionally independent given the class label
- B. Always independent (unconditionally)
- C. Always normally distributed
- D. Always binary

In the EM algorithm for latent-variable models, the E-step primarily computes:

- A. Expected latent-variable responsibilities given current parameters
- B. A new learning rate schedule
- C. The exact global optimum in one step
- D. A decision boundary that maximizes the margin