#### ML -Assignments - 22April 2025

#### 1. Decision Tree Classification

Use make\_classification(n\_samples=100, n\_features=2) to generate a binary classification dataset.

Train a DecisionTreeClassifier, predict, and print the classification report.

## 2. Decision Tree Regression

Generate data using make\_regression(n\_samples=100, n\_features=1, noise=10). Train a DecisionTreeRegressor and evaluate using MAE, RMSE, and R<sup>2</sup> score.

#### 3. Plot Decision Boundaries of a Decision Tree

Use a 2D dataset (make\_classification with 2 features), fit a DecisionTreeClassifier, and plot the **decision boundary**.

## 4. Max Depth Impact in Decision Trees

Use the Iris dataset.

Train DecisionTreeClassifier with different max\_depth values and compare accuracy on the test set.

#### 5. Visualize Tree Structure

Train a decision tree on the Iris dataset and use plot\_tree from sklearn.tree to visualize the structure of the tree.

#### Random Forest

#### 6. Random Forest Classification

Use the Breast Cancer dataset from sklearn.datasets.load\_breast\_cancer.

Train a RandomForestClassifier, compute accuracy, and display feature importance.

#### 7. Random Forest Regression

Create data using make regression.

Train a RandomForestRegressor, evaluate on test data using R<sup>2</sup> and RMSE, and visualize predicted vs actual values.

## 8. Compare Random Forest vs Decision Tree

On a classification dataset (make\_classification), compare accuracy, **F1-score**, and **confusion matrix** between a DecisionTreeClassifier and a RandomForestClassifier.

# Gradient Boosting Machines (GBM)

## 9. Gradient Boosting Classifier

Use the Wine dataset (load\_wine() from sklearn). Train a GradientBoostingClassifier and compare accuracy with RandomForestClassifier.

## 10. Gradient Boosting Regressor

Generate regression data and train GradientBoostingRegressor. Tune parameters like n estimators, learning rate and observe impact on  $\mathbf{R}^2$  score.

# Model Stacking

#### 11. Stacking Classifier

On a classification dataset, stack LogisticRegression, DecisionTreeClassifier, and RandomForestClassifier using sklearn.ensemble.StackingClassifier. Compare accuracy with individual base models.

#### 12. Stacking Regressor

On regression data, implement stacking using RandomForestRegressor, GradientBoostingRegressor, and LinearRegression.

Measure RMSE.

#### CatBoost

#### 13. CatBoost Classifier

Use a dataset with categorical features (e.g., Titanic dataset or simulate one with pandas). Train a CatBoostClassifier and evaluate accuracy and F1-score.

# LightGBM

## 14. LightGBM Classifier

Generate a dataset with 10+ features using make\_classification. Train a LGBMClassifier, plot feature importance, and compare it to RandomForestClassifier.

#### XGBoost

## 15. XGBoost Regressor

Use make regression() data.

Train an XGBRegressor, tune parameters ( $n_estimators$ ,  $max_depth$ , learning\_rate), and evaluate using  $R^2$  and RMSE.

#### 16. Compare Hierarchical Clustering with K-Means

Dataset: Use make blobs (n samples=300, centers=4).

- Cluster the dataset using both AgglomerativeClustering and KMeans.
- Compare the cluster assignments visually.

## 17. Cluster Real-World Dataset (e.g., Mall Customer Segmentation)

Dataset: Use the Mall Customers.csv dataset (Age, Income, Spending Score).

- Normalize features using StandardScaler.
- Apply AgglomerativeClustering.
- Visualize clusters using scatter plots.

## 18. Determine Optimal Number of Clusters Using Dendrogram

Dataset: make blobs(n\_samples=100, centers=4, random\_state=42)

• Plot the dendrogram and cut it at different heights to determine the number of clusters.

## 19. Perform Agglomerative Clustering and Visualize Dendrogram

Dataset: Generate data using make\_blobs(n\_samples=150, centers=3, cluster\_std=1.0).

- Apply AgglomerativeClustering from sklearn.
- Use scipy.cluster.hierarchy.dendrogram to visualize the hierarchical tree
- Plot clusters using different colors.
- 20. Use Breast cancer dataset and perform the above .