## 306281422 stats101c hw3

## November 10, 2024

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[]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     from sklearn import datasets
     from sklearn.model_selection import train_test_split
     from sklearn.metrics import accuracy score
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.ensemble import BaggingClassifier, RandomForestClassifier,
      → Gradient Boosting Classifier
     data = pd.read_csv("banknote.csv", header=None)
     dataset = np.array(data)
     X = dataset[:, 0:4]
     \#X_1 = (X[:, 0] - X[:, 0].mean()) / X[:, 0].std()
     \#X_2 = (X[:, 1] - X[:, 1].mean()) / X[:, 1].std()
     y = dataset[:, 4]
[]: # Split the data into train and test sets
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,_
      →random_state=22)
[]: # Step 1: Evaluate the base classifier (Decision Tree)
     dtree = DecisionTreeClassifier(max_depth=5, random_state=22)
     dtree.fit(X_train, y_train)
     # Predict and evaluate the Decision Tree model
     y_pred_dt = dtree.predict(X_test)
     print("Decision Tree")
     print("Train data accuracy:", accuracy_score(y_true=y_train, y_pred=dtree.
      →predict(X train)))
     print("Test data accuracy:", accuracy_score(y_true=y_test, y_pred=y_pred_dt))
     print("Test Error Rate:", 1 - accuracy_score(y_true=y_test, y_pred=y_pred dt))
    Decision Tree
    Train data accuracy: 0.9895833333333334
```

Test data accuracy: 0.9635922330097088

Test Error Rate: 0.036407766990291246

## Random Forest

Train data accuracy: 0.990625

Test data accuracy: 0.9951456310679612 Test Error Rate: 0.004854368932038833

## Gradient Boosting

Train data accuracy: 1.0

Test data accuracy: 0.9927184466019418 Test Error Rate: 0.007281553398058249

Both Random Forest and Boosting models have significantly low test error compared to Decision Tree model. This suggests that Random Forest and Boosting generalize better on the dataset.