

Untitled33

April 20, 2025

```
[21]: # Step 1: Load and Preview the Heart Disease dataset
import pandas as pd

try:
    df = pd.read_csv("simulated_heart_disease.csv")
    print(" Dataset loaded successfully!\n")
    print(df.head())

    print("\n Missing value counts:")
    print(df.isnull().sum())

except FileNotFoundError:
    print(" File not found. Make sure 'simulated_heart_disease.csv' is in the_
    ↪same folder as this notebook.")
```

Dataset loaded successfully!

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	\
0	67	1	2	126	458	1	0	144	0	2.3	2	
1	57	0	0	158	384	0	1	133	0	6.2	0	
2	43	0	3	111	286	0	0	130	0	2.8	1	
3	71	1	2	189	515	1	1	149	0	2.1	1	
4	36	0	0	142	303	0	0	107	1	3.6	1	

	ca	thal	target
0	2	2	1
1	4	0	0
2	0	2	0
3	2	1	1
4	0	0	1

Missing value counts:

age	0
sex	0
cp	0
trestbps	0
chol	0
fbs	0

```
restecg      0
thalach      0
exang        0
oldpeak      0
slope        0
ca           0
thal         0
target       0
dtype: int64
```

```
[22]: from sklearn.model_selection import train_test_split

# Split into features and target
X = df.drop("target", axis=1)
y = df["target"]

# Train/test split
X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.8,
↳ random_state=42)
print("Train shape:", X_train.shape)
print("Test shape:", X_test.shape)
```

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Train shape: (242, 13)
Test shape: (61, 13)
```

```
[23]: from sklearn.svm import SVC
from sklearn.ensemble import GradientBoostingClassifier, RandomForestClassifier

# Models + hyperparameter grids
grid_params = {
    'SVM': {
        'model': SVC(probability=True, random_state=42),
        'params': {'kernel': ['linear', 'rbf'], 'C': [0.1, 1, 10]}
    },
    'GradientBoosting': {
        'model': GradientBoostingClassifier(random_state=42),
        'params': {'n_estimators': [100, 150], 'learning_rate': [0.1, 0.05],
↳ 'max_depth': [3, 5]}
    },
    'RandomForest': {
        'model': RandomForestClassifier(random_state=42),
        'params': {'n_estimators': [100, 150], 'max_depth': [None, 10],
↳ 'min_samples_split': [2, 5]}
    }
}
```

```
[24]: from sklearn.model_selection import GridSearchCV
```

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from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, roc_auc_score

results = []
best_models = {}

for name, gp in grid_params.items():
    print(f"Training {name}...")
    clf = GridSearchCV(gp['model'], gp['params'], cv=5, scoring='f1')
    clf.fit(X_train, y_train)
    best = clf.best_estimator_
    best_models[name] = best

    y_pred = best.predict(X_test)
    y_proba = best.predict_proba(X_test)[: , 1]

    # Store metrics
    results.append({
        'Model': name,
        'Accuracy': round(accuracy_score(y_test, y_pred), 3),
        'Precision': round(precision_score(y_test, y_pred), 3),
        'Recall': round(recall_score(y_test, y_pred), 3),
        'F1-score': round(f1_score(y_test, y_pred), 3),
        'AUC-ROC': round(roc_auc_score(y_test, y_proba), 3)
    })

print(" All models trained.")

```

Training SVM...

Training GradientBoosting...

Training RandomForest...

All models trained.

```

[25]: import matplotlib.pyplot as plt
from sklearn.metrics import roc_curve, roc_auc_score

# Summary table
df_results = pd.DataFrame(results).set_index('Model')
print(" Model Performance Summary:")
display(df_results) # Use display() for better notebook output

# Plot ROC Curves
plt.figure(figsize=(8, 6))

for name, model in best_models.items():
    try:
        y_proba = model.predict_proba(X_test)[: , 1]

```

```

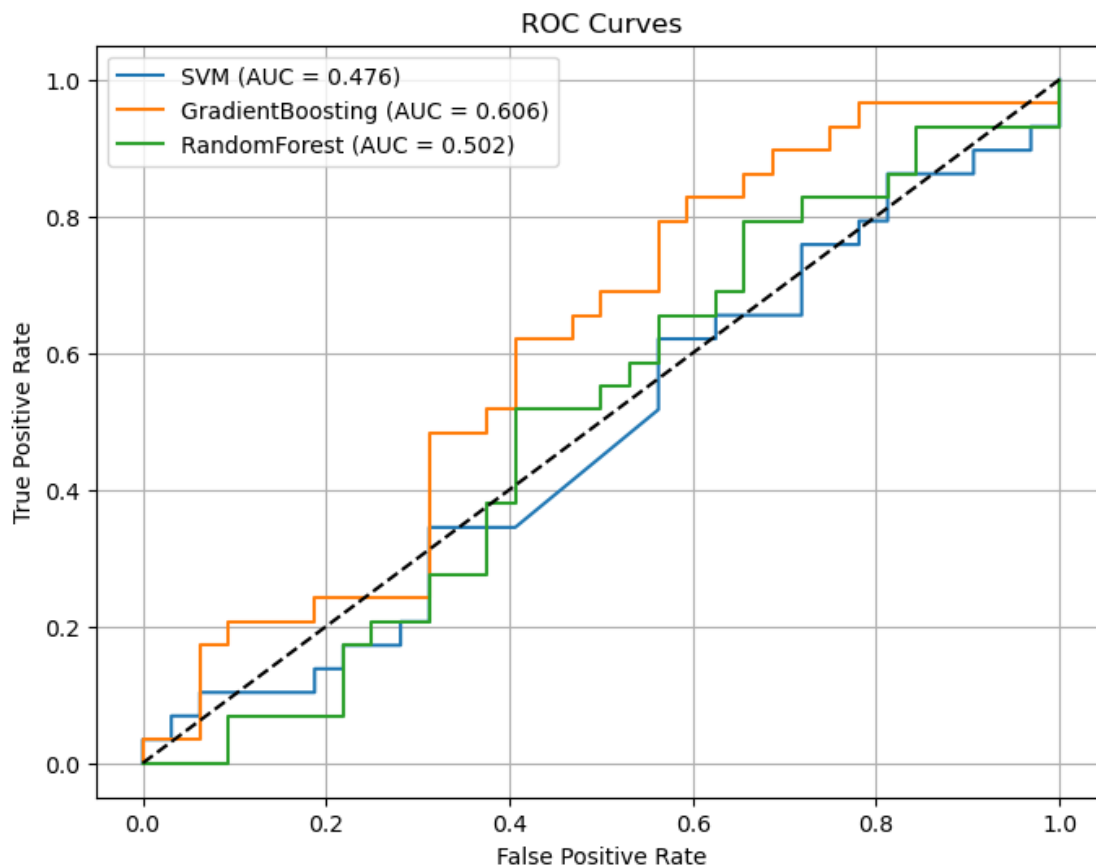
fpr, tpr, _ = roc_curve(y_test, y_proba)
auc = roc_auc_score(y_test, y_proba)
plt.plot(fpr, tpr, label=f"{name} (AUC = {round(auc, 3)})")
except Exception as e:
    print(f" Skipped {name} due to error: {e}")

plt.plot([0, 1], [0, 1], 'k--')
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ROC Curves")
plt.legend()
plt.grid(True)
plt.show()

```

Model Performance Summary:

	Accuracy	Precision	Recall	F1-score	AUC-ROC
Model					
SVM	0.475	0.452	0.483	0.467	0.476
GradientBoosting	0.590	0.567	0.586	0.576	0.606
RandomForest	0.525	0.500	0.448	0.473	0.502



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