

cation-diabetesdataset-16-11-24-1

November 24, 2024

1 Binary Classification

Problem Statement - The goal of this project is to build, evaluate, and compare three different models to classify whether a patient has diabetes.

Dataset Overview - We are using the Pima Indians Diabetes Dataset, which is a well-known dataset in machine learning for binary classification tasks. The dataset consists of medical diagnostic measurements to predict whether a patient has diabetes (Outcome = 1) or not (Outcome = 0). The dataset has 768 rows and 8 input features:

- Pregnancies: Number of times pregnant.
- Glucose: Plasma glucose concentration (mg/dL).
- BloodPressure: Diastolic blood pressure (mm Hg).
- SkinThickness: Triceps skinfold thickness (mm).
- Insulin: 2-hour serum insulin (μ U/ml).
- BMI: Body mass index ($\text{weight in kg}/(\text{height in m})^2$).
- DiabetesPedigreeFunction: A function that represents diabetes hereditary risk.
- Age: Age of the patient (years).
- Outcome: Target variable (0 = No diabetes, 1 = Diabetes).

1.0.1 Install all necessary packages

```
[1]: !pip install pandas
      !pip install numpy
      !pip install matplotlib
      !pip install seaborn
      !pip install scikit-learn
      !pip install tensorflow
      !pip install keras-models
```

Requirement already satisfied: pandas in /usr/local/lib/python3.10/dist-packages (2.2.2)

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 Requirement already satisfied: wheel<1.0,>=0.23.0 in
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 Requirement already satisfied: charset-normalizer<4,>=2 in
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 Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0 in

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Requirement already satisfied: optree in /usr/local/lib/python3.10/dist-packages
(from keras->keras-models) (0.13.1)
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packages (from keras->keras-models) (0.4.1)
Requirement already satisfied: packaging in /usr/local/lib/python3.10/dist-
packages (from keras->keras-models) (24.2)
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Requirement already satisfied: spacy-loggers<2.0.0,>=1.0.0 in
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Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in
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Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in
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Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in
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Requirement already satisfied: weasel<0.5.0,>=0.1.0 in
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Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in
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Requirement already satisfied: requests<3.0.0,>=2.13.0 in
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Requirement already satisfied: pydantic!=1.8,!1.8.1,<3.0.0,>=1.7.4 in
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 (from spacy->keras-models) (3.1.4)

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 packages (from spacy->keras-models) (75.1.0)

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models) (0.7.11)
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models) (0.1.5)
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/usr/local/lib/python3.10/dist-packages (from language-
data>=1.2->langcodes<4.0.0,>=3.2.0->spacy->keras-models) (1.2.1)
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packages (from markdown-it-py>=2.2.0->rich->keras->keras-models) (0.1.2)
Requirement already satisfied: wrapt in /usr/local/lib/python3.10/dist-packages
(from smart-open<8.0.0,>=5.2.1->weasel<0.5.0,>=0.1.0->spacy->keras-models)
(1.16.0)
Downloading keras_models-0.0.7-py3-none-any.whl (18 kB)
Installing collected packages: keras-models
Successfully installed keras-models-0.0.7

```

1.0.2 Import All Necessary Libraries & Packages

```

[2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

```

```

import seaborn as sns
from sklearn.model_selection import train_test_split, GridSearchCV, \
    StratifiedKFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import (
    confusion_matrix, roc_auc_score, roc_curve, classification_report, \
    log_loss, matthews_corrcoef
)
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, LSTM
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.utils import to_categorical

```

1.0.3 Load the dataset

```

[3]: # Load Dataset
data = pd.read_csv("diabetes.csv")
data.head()

```

```

[3]:   Pregnancies  Glucose  BloodPressure  SkinThickness  Insulin   BMI   \
0             6      148                72             35         0  33.6
1             1       85                66             29         0  26.6
2             8      183                64              0         0  23.3
3             1       89                66             23        94  28.1
4             0      137                40             35       168  43.1

      DiabetesPedigreeFunction  Age  Outcome
0                0.627      50         1
1                0.351      31         0
2                0.672      32         1
3                0.167      21         0
4                2.288      33         1

```

Basic Information on Dataset

```

[4]: # Display Shape and Info
print("Dataset Shape:", data.shape)
print("\nDataset Info:")
print(data.info())

```

Dataset Shape: (768, 9)

Dataset Info:

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767

```


Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	Pregnancies	768 non-null	int64
1	Glucose	768 non-null	int64
2	BloodPressure	768 non-null	int64
3	SkinThickness	768 non-null	int64
4	Insulin	768 non-null	int64
5	BMI	768 non-null	float64
6	DiabetesPedigreeFunction	768 non-null	float64
7	Age	768 non-null	int64
8	Outcome	768 non-null	int64

dtypes: float64(2), int64(7)

memory usage: 54.1 KB

None

1.0.4 EDA

```
[5]: # Basic EDA
print("\nNull Values:\n", data.isnull().sum())
data.drop_duplicates(inplace=True)
print("\nAfter Removing Duplicates - Shape:", data.shape)
```

Null Values:

Pregnancies	0
Glucose	0
BloodPressure	0
SkinThickness	0
Insulin	0
BMI	0
DiabetesPedigreeFunction	0
Age	0
Outcome	0

dtype: int64

After Removing Duplicates - Shape: (768, 9)

Describe Dataset

```
[6]: # Describe Dataset
print("\nDataset Description:")
print(data.describe())
```

Dataset Description:

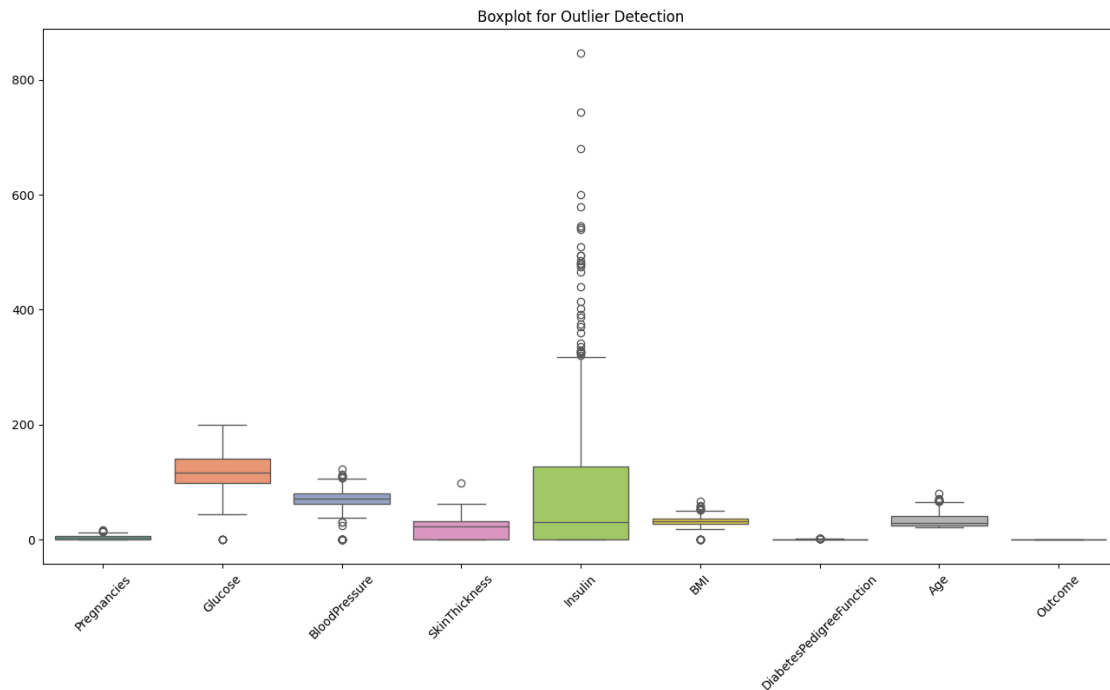
	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	\
count	768.000000	768.000000	768.000000	768.000000	768.000000	
mean	3.845052	120.894531	69.105469	20.536458	79.799479	

std	3.369578	31.972618	19.355807	15.952218	115.244002
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	1.000000	99.000000	62.000000	0.000000	0.000000
50%	3.000000	117.000000	72.000000	23.000000	30.500000
75%	6.000000	140.250000	80.000000	32.000000	127.250000
max	17.000000	199.000000	122.000000	99.000000	846.000000

	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000
mean	31.992578	0.471876	33.240885	0.348958
std	7.884160	0.331329	11.760232	0.476951
min	0.000000	0.078000	21.000000	0.000000
25%	27.300000	0.243750	24.000000	0.000000
50%	32.000000	0.372500	29.000000	0.000000
75%	36.600000	0.626250	41.000000	1.000000
max	67.100000	2.420000	81.000000	1.000000

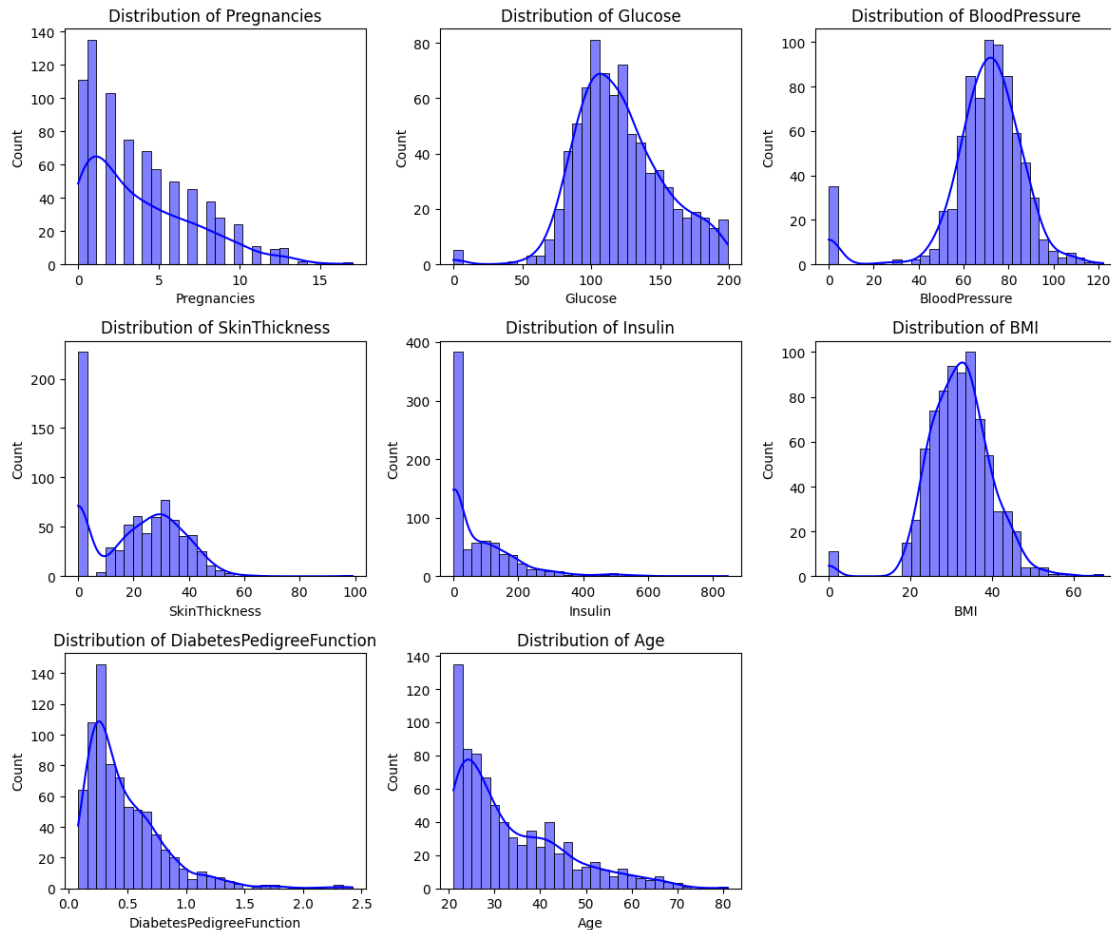
Outlier Detection

```
[7]: # Detect Outliers with Boxplot
plt.figure(figsize=(16, 8))
sns.boxplot(data=data, palette="Set2")
plt.title("Boxplot for Outlier Detection")
plt.xticks(rotation=45)
plt.show()
```



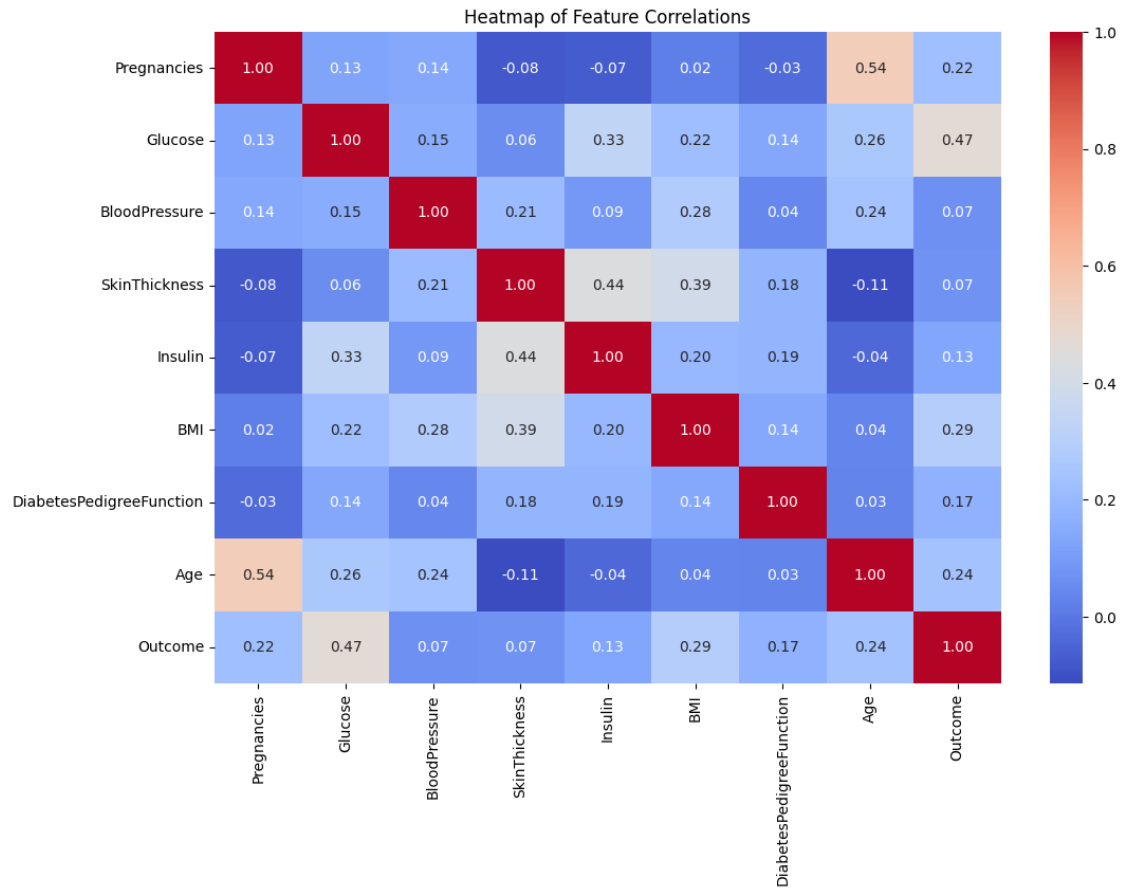
Some Basic Plots

```
[8]: # Histograms for Feature Distribution
plt.figure(figsize=(12, 10))
for i, column in enumerate(data.columns[:-1], start=1): # Exclude Outcome
    plt.subplot(3, 3, i)
    sns.histplot(data[column], kde=True, color='blue', bins=30)
    plt.title(f'Distribution of {column}')
plt.tight_layout()
plt.show()
```



Correlation Heatmap

```
[9]: # Heatmap for Feature Correlations
plt.figure(figsize=(12, 8))
sns.heatmap(data.corr(), annot=True, cmap="coolwarm", fmt=".2f")
plt.title("Heatmap of Feature Correlations")
plt.show()
```



1.0.5 Standardization

```
[10]: # Preprocessing (Standardization)
scaler = StandardScaler()
X = scaler.fit_transform(data.iloc[:, :-1]) # Features
y = data['Outcome'] # Target
```

1.0.6 Machine Learning Models

1.0.7 GridSearchCV - Finding Best Hyperparameter

```
[11]: rf_params = {
    'n_estimators': [50, 100, 150],
    'max_depth': [3, 5, 10],
    'min_samples_split': [2, 5, 10]
}
rf_grid = GridSearchCV(RandomForestClassifier(random_state=42), rf_params,
    cv=5, scoring='accuracy')
rf_grid.fit(X, y)
```

```
print("Best RF Hyperparameters:", rf_grid.best_params_)
```

Best RF Hyperparameters: {'max_depth': 5, 'min_samples_split': 2, 'n_estimators': 100}

/usr/local/lib/python3.10/dist-packages/numpy/ma/core.py:2820: RuntimeWarning:
invalid value encountered in cast

```
_data = np.array(data, dtype=dtype, copy=copy,
```

```
[12]: # Model Training - SVM
svm_params = {
    'C': [0.1, 1, 10],
    'kernel': ['linear', 'rbf'],
    'gamma': ['scale', 'auto']
}
svm_grid = GridSearchCV(SVC(probability=True), svm_params, cv=5,
    <scoring='accuracy')
svm_grid.fit(X, y)
print("Best SVM Hyperparameters:", svm_grid.best_params_)
```

Best SVM Hyperparameters: {'C': 1, 'gamma': 'scale', 'kernel': 'linear'}

Cross-Validation & Model Training

```
[13]: def train_ml_model(model, X, y, model_name):
    skf = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
    metrics = []

    print(f"\nEvaluating {model_name} with 10-Fold Cross-Validation:")
    for fold, (train_index, test_index) in enumerate(skf.split(X, y), start=1):
        X_train, X_test = X[train_index], X[test_index]
        y_train, y_test = y[train_index], y[test_index]

        # Fit and Predict
        model.fit(X_train, y_train)
        y_pred = model.predict(X_test)
        y_prob = model.predict_proba(X_test)[: , 1]

        # Confusion Matrix
        cm = confusion_matrix(y_test, y_pred)
        tn, fp, fn, tp = cm.ravel()

        # Metrics Calculation
        tpr = tp / (tp + fn)
        tnr = tn / (tn + fp)
        fpr = fp / (tn + fp)
        fnr = fn / (tp + fn)
        precision = tp / (tp + fp)
```

```

f1 = 2 * tp / (2 * tp + fp + fn)
accuracy = (tp + tn) / (tp + fp + fn + tn)
error_rate = (fp + fn) / (tp + fp + fn + tn)
bacc = (tpr + tnr) / 2
tss = tpr - fpr
hss = 2 * (tp * tn - fp * fn) / ((tp + fn) * (fn + tn) + (tp + fp) *
↪(fp + tn))
recall = tp / tp + fn

fold_metrics = {
    'Fold': fold,
    'TP': tp, 'TN': tn, 'FP': fp, 'FN': fn,
    "TPR": tpr, "TNR": tnr, "FPR": fpr, "FNR": fnr,
    'Accuracy': accuracy,
    'Precision': precision,
    'Recall': recall,
    'F1 Score': f1,
    "bacc": bacc,
    "tss": tss,
    "hss": hss,

}

metrics.append(fold_metrics)

# Print Metrics for the Current Fold
print(f"Fold {fold} Metrics:")
for metric, value in fold_metrics.items():
    print(f"{metric}: {value}")
print("_"*30+"\n")

return model, pd.DataFrame(metrics)

```

1.0.8 Random Forest Model Training

```

[14]: # Evaluate Random Forest
best_rf, rf_metrics = train_ml_model(
    RandomForestClassifier(**rf_grid.best_params_), X, y, "Random Forest"
)

```

Evaluating Random Forest with 10-Fold Cross-Validation:

Fold 1 Metrics:

Fold: 1

TP: 18

TN: 45

FP: 5

FN: 9

TPR: 0.6666666666666666
TNR: 0.9
FPR: 0.1
FNR: 0.3333333333333333
Accuracy: 0.8181818181818182
Precision: 0.782608695652174
Recall: 10.0
F1 Score: 0.72
bacc: 0.7833333333333333
tss: 0.5666666666666667
hss: 0.5866564417177914

Fold 2 Metrics:

Fold: 2
TP: 13
TN: 46
FP: 4
FN: 14
TPR: 0.48148148148148145
TNR: 0.92
FPR: 0.08
FNR: 0.5185185185185185
Accuracy: 0.7662337662337663
Precision: 0.7647058823529411
Recall: 15.0
F1 Score: 0.5909090909090909
bacc: 0.7007407407407408
tss: 0.40148148148148144
hss: 0.4388663967611336

Fold 3 Metrics:

Fold: 3
TP: 14
TN: 43
FP: 7
FN: 13
TPR: 0.5185185185185185
TNR: 0.86
FPR: 0.14
FNR: 0.48148148148148145
Accuracy: 0.7402597402597403
Precision: 0.6666666666666666
Recall: 14.0
F1 Score: 0.5833333333333334
bacc: 0.6892592592592592
tss: 0.3785185185185185

hss: 0.3989071038251366

Fold 4 Metrics:

Fold: 4

TP: 18

TN: 49

FP: 1

FN: 9

TPR: 0.6666666666666666

TNR: 0.98

FPR: 0.02

FNR: 0.3333333333333333

Accuracy: 0.8701298701298701

Precision: 0.9473684210526315

Recall: 10.0

F1 Score: 0.782608695652174

bacc: 0.8233333333333333

tss: 0.6466666666666666

hss: 0.6939586645468998

Fold 5 Metrics:

Fold: 5

TP: 18

TN: 47

FP: 3

FN: 9

TPR: 0.6666666666666666

TNR: 0.94

FPR: 0.06

FNR: 0.3333333333333333

Accuracy: 0.8441558441558441

Precision: 0.8571428571428571

Recall: 10.0

F1 Score: 0.75

bacc: 0.8033333333333332

tss: 0.6066666666666667

hss: 0.639344262295082

Fold 6 Metrics:

Fold: 6

TP: 14

TN: 46

FP: 4

FN: 13

TPR: 0.5185185185185185

TNR: 0.92
FPR: 0.08
FNR: 0.48148148148148145
Accuracy: 0.7792207792207793
Precision: 0.7777777777777778
Recall: 14.0
F1 Score: 0.6222222222222222
bacc: 0.7192592592592593
tss: 0.4385185185185185
hss: 0.47492980344965907

Fold 7 Metrics:

Fold: 7
TP: 12
TN: 45
FP: 5
FN: 15
TPR: 0.4444444444444444
TNR: 0.9
FPR: 0.1
FNR: 0.5555555555555556
Accuracy: 0.7402597402597403
Precision: 0.7058823529411765
Recall: 16.0
F1 Score: 0.5454545454545454
bacc: 0.6722222222222223
tss: 0.3444444444444444
hss: 0.3765182186234818

Fold 8 Metrics:

Fold: 8
TP: 15
TN: 41
FP: 9
FN: 12
TPR: 0.5555555555555556
TNR: 0.82
FPR: 0.18
FNR: 0.4444444444444444
Accuracy: 0.7272727272727273
Precision: 0.625
Recall: 13.0
F1 Score: 0.5882352941176471
bacc: 0.6877777777777778
tss: 0.3755555555555556
hss: 0.38540478905359177

```
-----  
Fold 9 Metrics:  
Fold: 9  
TP: 16  
TN: 40  
FP: 10  
FN: 10  
TPR: 0.6153846153846154  
TNR: 0.8  
FPR: 0.2  
FNR: 0.38461538461538464  
Accuracy: 0.7368421052631579  
Precision: 0.6153846153846154  
Recall: 11.0  
F1 Score: 0.6153846153846154  
bacc: 0.7076923076923077  
tss: 0.4153846153846154  
hss: 0.4153846153846154  
-----
```

```
Fold 10 Metrics:  
Fold: 10  
TP: 12  
TN: 44  
FP: 6  
FN: 14  
TPR: 0.46153846153846156  
TNR: 0.88  
FPR: 0.12  
FNR: 0.5384615384615384  
Accuracy: 0.7368421052631579  
Precision: 0.6666666666666666  
Recall: 15.0  
F1 Score: 0.5454545454545454  
bacc: 0.6707692307692308  
tss: 0.34153846153846157  
hss: 0.3687707641196013  
-----
```

1.0.9 SVM Model Training

```
[15]: # Evaluate SVM  
best_svm, svm_metrics = train_ml_model(  
    SVC(probability=True, **svm_grid.best_params_), X, y, "SVM"  
)
```

Evaluating SVM with 10-Fold Cross-Validation:

Fold 1 Metrics:

Fold: 1

TP: 16

TN: 44

FP: 6

FN: 11

TPR: 0.5925925925925926

TNR: 0.88

FPR: 0.12

FNR: 0.4074074074074074

Accuracy: 0.7792207792207793

Precision: 0.7272727272727273

Recall: 12.0

F1 Score: 0.6530612244897959

bacc: 0.7362962962962962

tss: 0.47259259259259256

hss: 0.49361702127659574

Fold 2 Metrics:

Fold: 2

TP: 14

TN: 45

FP: 5

FN: 13

TPR: 0.5185185185185185

TNR: 0.9

FPR: 0.1

FNR: 0.48148148148148145

Accuracy: 0.7662337662337663

Precision: 0.7368421052631579

Recall: 14.0

F1 Score: 0.6086956521739131

bacc: 0.7092592592592593

tss: 0.4185185185185185

hss: 0.4491255961844197

Fold 3 Metrics:

Fold: 3

TP: 15

TN: 43

FP: 7

FN: 12

TPR: 0.5555555555555556

TNR: 0.86

FPR: 0.14
FNR: 0.4444444444444444
Accuracy: 0.7532467532467533
Precision: 0.6818181818181818
Recall: 13.0
F1 Score: 0.6122448979591837
bacc: 0.7077777777777778
tss: 0.4155555555555557
hss: 0.4340425531914894

Fold 4 Metrics:

Fold: 4
TP: 15
TN: 48
FP: 2
FN: 12
TPR: 0.5555555555555556
TNR: 0.96
FPR: 0.04
FNR: 0.4444444444444444
Accuracy: 0.8181818181818182
Precision: 0.8823529411764706
Recall: 13.0
F1 Score: 0.6818181818181818
bacc: 0.7577777777777778
tss: 0.5155555555555555
hss: 0.5635627530364372

Fold 5 Metrics:

Fold: 5
TP: 15
TN: 47
FP: 3
FN: 12
TPR: 0.5555555555555556
TNR: 0.94
FPR: 0.06
FNR: 0.4444444444444444
Accuracy: 0.8051948051948052
Precision: 0.8333333333333334
Recall: 13.0
F1 Score: 0.6666666666666666
bacc: 0.7477777777777778
tss: 0.4955555555555556
hss: 0.5367027677496992

Fold 6 Metrics:

Fold: 6

TP: 17

TN: 44

FP: 6

FN: 10

TPR: 0.6296296296296297

TNR: 0.88

FPR: 0.12

FNR: 0.37037037037037035

Accuracy: 0.7922077922077922

Precision: 0.7391304347826086

Recall: 11.0

F1 Score: 0.68

bacc: 0.7548148148148148

tss: 0.5096296296296297

hss: 0.5276073619631901

Fold 7 Metrics:

Fold: 7

TP: 13

TN: 47

FP: 3

FN: 14

TPR: 0.48148148148148145

TNR: 0.94

FPR: 0.06

FNR: 0.5185185185185185

Accuracy: 0.7792207792207793

Precision: 0.8125

Recall: 15.0

F1 Score: 0.6046511627906976

bacc: 0.7107407407407407

tss: 0.42148148148148146

hss: 0.46505925623212097

Fold 8 Metrics:

Fold: 8

TP: 18

TN: 39

FP: 11

FN: 9

TPR: 0.6666666666666666

TNR: 0.78

FPR: 0.22

FNR: 0.3333333333333333
Accuracy: 0.7402597402597403
Precision: 0.6206896551724138
Recall: 10.0
F1 Score: 0.6428571428571429
bacc: 0.7233333333333334
tss: 0.4466666666666666
hss: 0.43918426802621996

Fold 9 Metrics:

Fold: 9
TP: 19
TN: 38
FP: 12
FN: 7
TPR: 0.7307692307692307
TNR: 0.76
FPR: 0.24
FNR: 0.2692307692307692
Accuracy: 0.75
Precision: 0.6129032258064516
Recall: 8.0
F1 Score: 0.6666666666666666
bacc: 0.7453846153846153
tss: 0.49076923076923074
hss: 0.46911764705882353

Fold 10 Metrics:

Fold: 10
TP: 13
TN: 44
FP: 6
FN: 13
TPR: 0.5
TNR: 0.88
FPR: 0.12
FNR: 0.5
Accuracy: 0.75
Precision: 0.6842105263157895
Recall: 14.0
F1 Score: 0.5777777777777777
bacc: 0.69
tss: 0.38
hss: 0.40625

1.0.10 Deep Learning Model - LSTM

```
[16]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, LSTM
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.utils import to_categorical
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import confusion_matrix, roc_auc_score, roc_curve, \
    log_loss, matthews_corrcoef

def train_lstm_model(X, y, input_shape, model_name="LSTM"):
    skf = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
    metrics = []

    print(f"\nEvaluating {model_name} with 10-Fold Cross-Validation:")

    for fold, (train_index, test_index) in enumerate(skf.split(X, y), start=1):
        # Prepare train and test sets
        X_train, X_test = X[train_index], X[test_index]
        y_train, y_test = y[train_index], y[test_index]

        # Reshape for LSTM
        X_train = X_train.reshape((X_train.shape[0], input_shape[0], \
    input_shape[1]))
        X_test = X_test.reshape((X_test.shape[0], input_shape[0], \
    input_shape[1]))
        y_train = to_categorical(y_train)
        y_test_bin = to_categorical(y_test)

        # Build LSTM model
        model = Sequential([
            LSTM(128, activation='relu', input_shape=(input_shape[0], \
    input_shape[1])),
            Dropout(0.2),
            Dense(2, activation='softmax')
        ])
        model.compile(optimizer=Adam(learning_rate=0.001), \
    loss='binary_crossentropy', metrics=['accuracy'])

        # Train model
        model.fit(X_train, y_train, epochs=10, verbose=0, batch_size=32)

        # Evaluate model
        y_pred_prob = model.predict(X_test)[:, 1]
        y_pred = (y_pred_prob > 0.5).astype(int)

        cm = confusion_matrix(y_test, y_pred)
```

```

tn, fp, fn, tp = cm.ravel()

# Metrics Calculation
tpr = tp / (tp + fn)
tnr = tn / (tn + fp)
fpr = fp / (tn + fp)
fnr = fn / (tp + fn)
precision = tp / (tp + fp)
f1 = 2 * tp / (2 * tp + fp + fn)
accuracy = (tp + tn) / (tp + fp + fn + tn)
error_rate = (fp + fn) / (tp + fp + fn + tn)
bacc = (tpr + tnr) / 2
tss = tpr - fpr
hss = 2 * (tp * tn - fp * fn) / ((tp + fn) * (fn + tn) + (tp + fp) *
↪(fp + tn))
recall = tp / tp + fn

fold_metrics = {
    'Fold': fold,
    'TP': tp, 'TN': tn, 'FP': fp, 'FN': fn,
    "TPR": tpr, "TNR": tnr, "FPR": fpr, "FNR": fnr,
    'Accuracy': accuracy,
    'Precision': precision,
    'Recall': recall,
    'F1 Score': f1,
    "bacc": bacc,
    "tss": tss,
    "hss": hss,

}

metrics.append(fold_metrics)

# Print Metrics for Current Fold
print(f"Fold {fold} Metrics:")
for metric, value in fold_metrics.items():
    print(f"{metric}: {value}")
print("_"*30+"\n")

return model, pd.DataFrame(metrics)

# Preprocessing: Scale and Reshape for LSTM
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
input_shape = (X_scaled.shape[1], 1) # Each feature is treated as a time step
X_scaled = X_scaled.reshape((X_scaled.shape[0], input_shape[0], input_shape[1]))

```



```
# Evaluate LSTM
best_lstm, lstm_metrics = train_lstm_model(X_scaled, y.values, input_shape)
```

Evaluating LSTM with 10-Fold Cross-Validation:

```
/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.
```

```
    super().__init__(**kwargs)
```

3/3 0s 75ms/step

Fold 1 Metrics:

Fold: 1

TP: 18

TN: 39

FP: 11

FN: 9

TPR: 0.6666666666666666

TNR: 0.78

FPR: 0.22

FNR: 0.3333333333333333

Accuracy: 0.7402597402597403

Precision: 0.6206896551724138

Recall: 10.0

F1 Score: 0.6428571428571429

bacc: 0.7233333333333334

tss: 0.4466666666666666

hss: 0.43918426802621996

```
/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
```

```
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.
```

```
    super().__init__(**kwargs)
```

3/3 0s 73ms/step

Fold 2 Metrics:

Fold: 2

TP: 12

TN: 43

FP: 7

FN: 15

TPR: 0.4444444444444444

TNR: 0.86

FPR: 0.14

FNR: 0.5555555555555556
Accuracy: 0.7142857142857143
Precision: 0.631578947368421
Recall: 16.0
F1 Score: 0.5217391304347826
bacc: 0.6522222222222223
tss: 0.3044444444444444
hss: 0.3267090620031797

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.

```
    super().__init__(**kwargs)
```

WARNING:tensorflow:5 out of the last 7 calls to <function
TensorFlowTrainer.make_predict_function.<locals>.one_step_on_data_distributed at
0x7d11fcb16c0> triggered tf.function retracing. Tracing is expensive and the
excessive number of tracings could be due to (1) creating @tf.function
repeatedly in a loop, (2) passing tensors with different shapes, (3) passing
Python objects instead of tensors. For (1), please define your @tf.function
outside of the loop. For (2), @tf.function has reduce_retracing=True option that
can avoid unnecessary retracing. For (3), please refer to
https://www.tensorflow.org/guide/function#controlling_retracing and
https://www.tensorflow.org/api_docs/python/tf/function for more details.

1/3 0s 155ms/step

WARNING:tensorflow:6 out of the last 9 calls to <function
TensorFlowTrainer.make_predict_function.<locals>.one_step_on_data_distributed at
0x7d11fcb16c0> triggered tf.function retracing. Tracing is expensive and the
excessive number of tracings could be due to (1) creating @tf.function
repeatedly in a loop, (2) passing tensors with different shapes, (3) passing
Python objects instead of tensors. For (1), please define your @tf.function
outside of the loop. For (2), @tf.function has reduce_retracing=True option that
can avoid unnecessary retracing. For (3), please refer to
https://www.tensorflow.org/guide/function#controlling_retracing and
https://www.tensorflow.org/api_docs/python/tf/function for more details.

3/3 0s 79ms/step

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.

```
    super().__init__(**kwargs)
```

Fold 3 Metrics:
Fold: 3
TP: 12

TN: 44
FP: 6
FN: 15
TPR: 0.4444444444444444
TNR: 0.88
FPR: 0.12
FNR: 0.5555555555555556
Accuracy: 0.7272727272727273
Precision: 0.6666666666666666
Recall: 16.0
F1 Score: 0.5333333333333333
bacc: 0.6622222222222223
tss: 0.3244444444444444
hss: 0.35138387484957884

3/3 0s 77ms/step
Fold 4 Metrics:
Fold: 4
TP: 18
TN: 46
FP: 4
FN: 9
TPR: 0.6666666666666666
TNR: 0.92
FPR: 0.08
FNR: 0.3333333333333333
Accuracy: 0.8311688311688312
Precision: 0.8181818181818182
Recall: 10.0
F1 Score: 0.7346938775510204
bacc: 0.7933333333333333
tss: 0.5866666666666667
hss: 0.6127659574468085

`/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:`
UserWarning: Do not pass an ``input_shape`/`input_dim`` argument to a layer. When using Sequential models, prefer using an ``Input(shape)`` object as the first layer in the model instead.

`super().__init__(**kwargs)`

3/3 1s 299ms/step
Fold 5 Metrics:
Fold: 5
TP: 18
TN: 37
FP: 13

FN: 9
TPR: 0.6666666666666666
TNR: 0.74
FPR: 0.26
FNR: 0.3333333333333333
Accuracy: 0.7142857142857143
Precision: 0.5806451612903226
Recall: 10.0
F1 Score: 0.6206896551724138
bacc: 0.7033333333333334
tss: 0.4066666666666666
hss: 0.39326647564469913

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.

super().__init__(**kwargs)

3/3 0s 76ms/step

Fold 6 Metrics:

Fold: 6

TP: 14

TN: 41

FP: 9

FN: 13

TPR: 0.5185185185185185

TNR: 0.82

FPR: 0.18

FNR: 0.48148148148148145

Accuracy: 0.7142857142857143

Precision: 0.6086956521739131

Recall: 14.0

F1 Score: 0.56

bacc: 0.6692592592592592

tss: 0.3385185185185185

hss: 0.3504601226993865

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.

super().__init__(**kwargs)

3/3 0s 118ms/step

Fold 7 Metrics:

Fold: 7
TP: 12
TN: 44
FP: 6
FN: 15
TPR: 0.4444444444444444
TNR: 0.88
FPR: 0.12
FNR: 0.5555555555555556
Accuracy: 0.7272727272727273
Precision: 0.6666666666666666
Recall: 16.0
F1 Score: 0.5333333333333333
bacc: 0.6622222222222223
tss: 0.3244444444444444
hss: 0.35138387484957884

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.

super().__init__(**kwargs)

3/3 0s 77ms/step
Fold 8 Metrics:
Fold: 8
TP: 13
TN: 38
FP: 12
FN: 14
TPR: 0.48148148148148145
TNR: 0.76
FPR: 0.24
FNR: 0.5185185185185185
Accuracy: 0.6623376623376623
Precision: 0.52
Recall: 15.0
F1 Score: 0.5
bacc: 0.6207407407407407
tss: 0.24148148148148146
hss: 0.24566691785983422

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first

```
layer in the model instead.  
    super().__init__(**kwargs)
```

```
3/3                0s 82ms/step  
Fold 9 Metrics:  
Fold: 9  
TP: 18  
TN: 37  
FP: 13  
FN: 8  
TPR: 0.6923076923076923  
TNR: 0.74  
FPR: 0.26  
FNR: 0.3076923076923077  
Accuracy: 0.7236842105263158  
Precision: 0.5806451612903226  
Recall: 9.0  
F1 Score: 0.631578947368421  
bacc: 0.7161538461538461  
tss: 0.4323076923076923  
hss: 0.41323529411764703
```

```
-----  
  
/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:  
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.
```

```
    super().__init__(**kwargs)  
  
3/3                0s 75ms/step  
Fold 10 Metrics:  
Fold: 10  
TP: 14  
TN: 35  
FP: 15  
FN: 12  
TPR: 0.5384615384615384  
TNR: 0.7  
FPR: 0.3  
FNR: 0.46153846153846156  
Accuracy: 0.6447368421052632  
Precision: 0.4827586206896552  
Recall: 13.0  
F1 Score: 0.509090909090909  
bacc: 0.6192307692307693  
tss: 0.23846153846153845  
hss: 0.23203592814371257
```

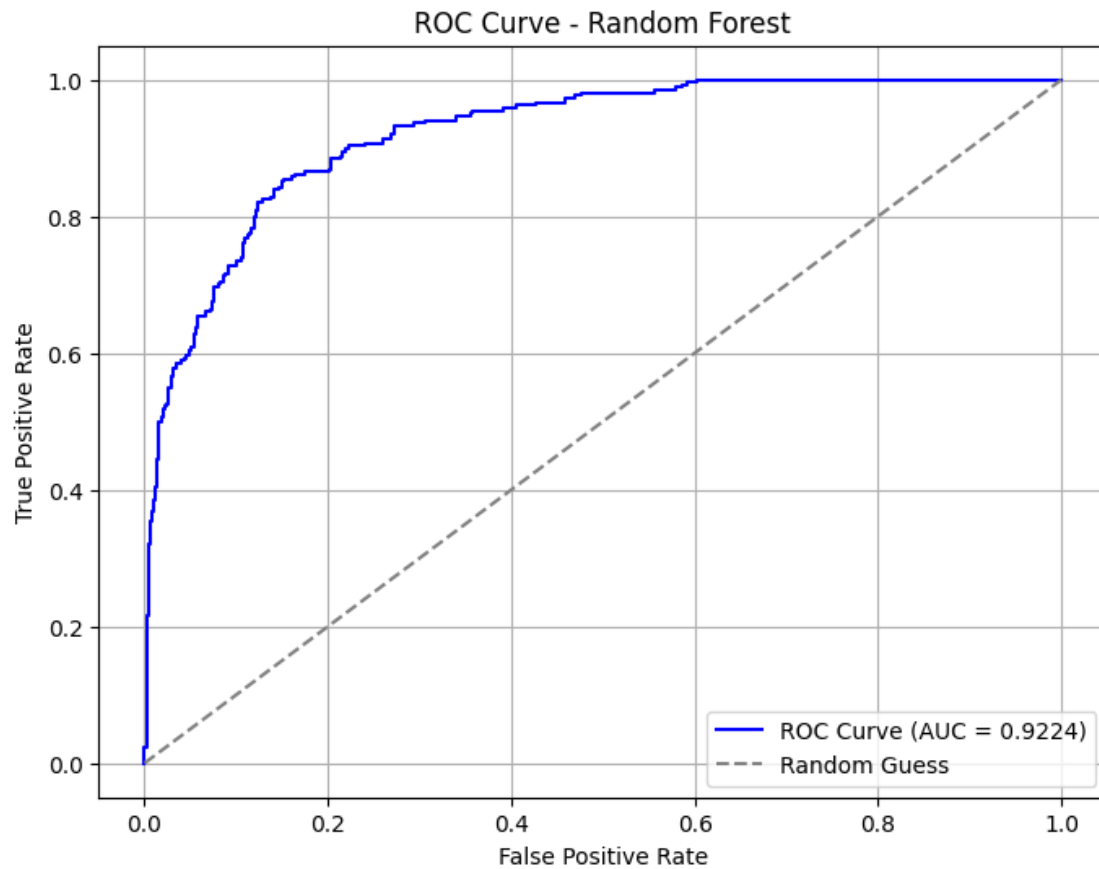
1.0.11 ROC Curve

```
[17]: import matplotlib.pyplot as plt
      from sklearn.metrics import roc_curve, auc

      # Function to plot ROC Curve
      def plot_roc_curve(y_test, y_proba, model_name):
          fpr, tpr, _ = roc_curve(y_test, y_proba)
          roc_auc = auc(fpr, tpr)
          plt.figure(figsize=(8, 6))
          plt.plot(fpr, tpr, color='blue', label=f'ROC Curve (AUC = {roc_auc:.4f})')
          plt.plot([0, 1], [0, 1], color='gray', linestyle='--', label='Random Guess')
          plt.title(f'ROC Curve - {model_name}')
          plt.xlabel('False Positive Rate')
          plt.ylabel('True Positive Rate')
          plt.legend(loc='lower right')
          plt.grid()
          plt.show()

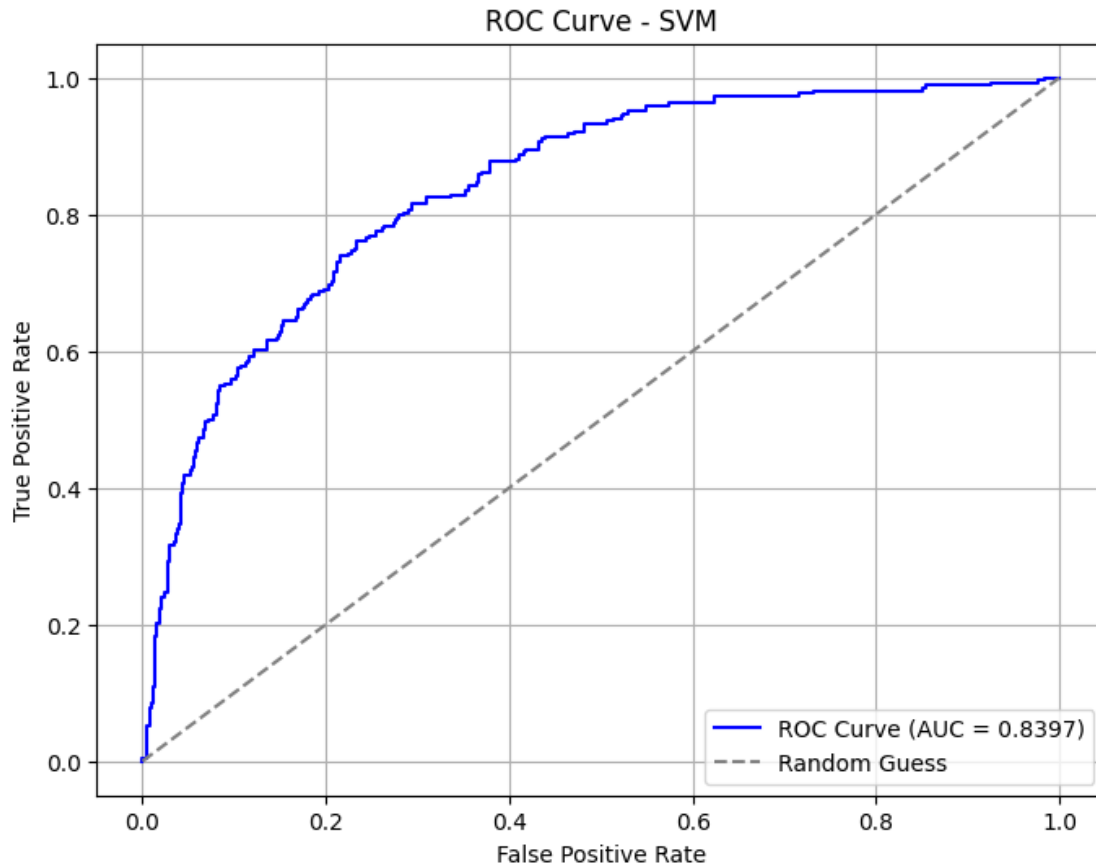
[18]: # Plot ROC Curve for Random Forest
      print("\nRandom Forest ROC Curve:")
      rf_y_proba = best_rf.predict_proba(X)[: , 1]
      plot_roc_curve(y, rf_y_proba, 'Random Forest')
```

Random Forest ROC Curve:



```
[19]: # Plot ROC Curve for SVM
print("\nSVM ROC Curve:")
svm_y_proba = best_svm.predict_proba(X)[: , 1]
plot_roc_curve(y, svm_y_proba, 'SVM')
```

SVM ROC Curve:



```
[20]: # Plot ROC Curve for LSTM
print("\nLSTM ROC Curve:")
X_dl = np.expand_dims(X, axis=2)
lstm_model = Sequential([
    LSTM(64, input_shape=(X_dl.shape[1], 1), return_sequences=False),
    Dropout(0.3),
    Dense(32, activation='relu'),
    Dense(1, activation='sigmoid')
])
lstm_model.compile(optimizer='adam', loss='binary_crossentropy',
    metrics=['accuracy'])
lstm_model.fit(X_dl, y, epochs=10, batch_size=16, verbose=0)
lstm_y_proba = lstm_model.predict(X_dl).flatten()
plot_roc_curve(y, lstm_y_proba, 'LSTM')
```

LSTM ROC Curve:

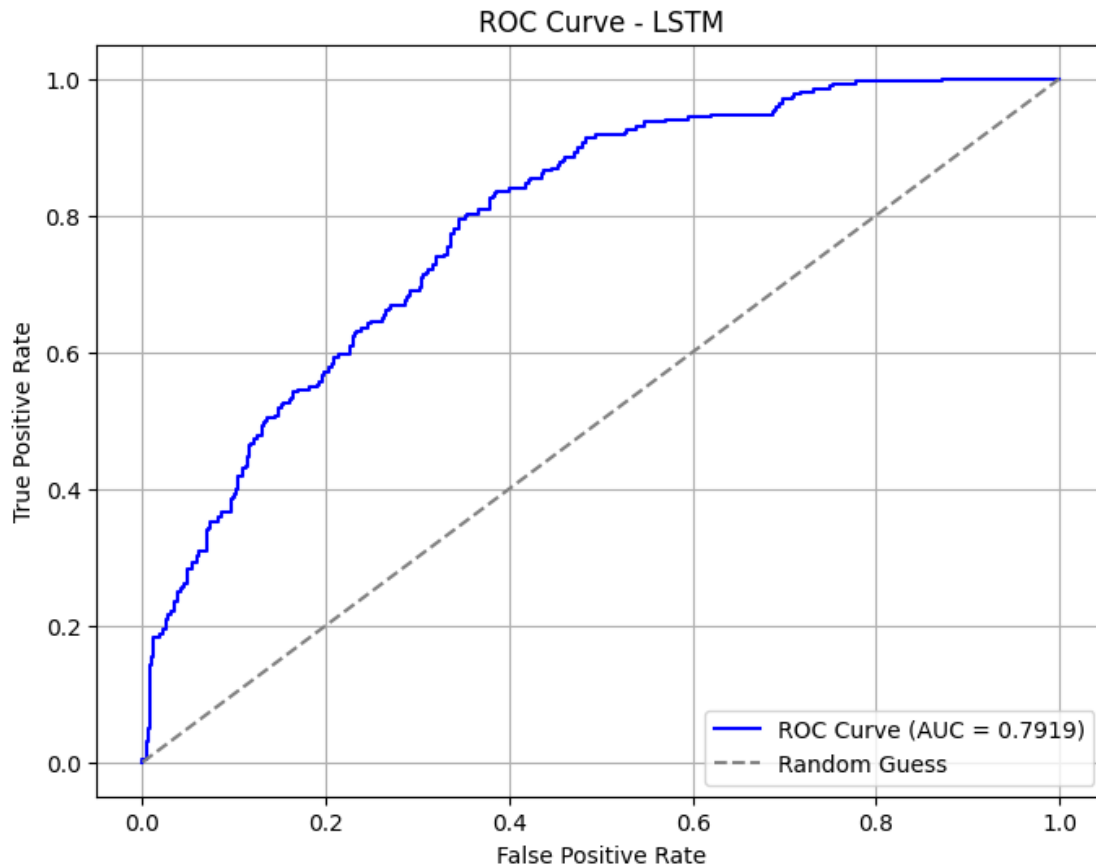
/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
 UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When

using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(**kwargs)
```

24/24

0s 3ms/step



1.0.12 Model Comparison

```
[22]: # Combine Metrics DataFrames
rf_metrics['Model'] = 'Random Forest'
svm_metrics['Model'] = 'SVM'
lstm_metrics['Model'] = 'LSTM'

comparison_df = pd.concat([rf_metrics, svm_metrics, lstm_metrics])
comparison_df = comparison_df[comparison_df['Fold']==10]
columns_to_display = ['Model', 'TP', 'TN', 'FP', 'FN',
                      "TPR", "TNR", "FPR", "FNR",
                      "Accuracy", "Precision", "Recall", "F1 Score",
                      "bacc", "tss", "hss",]
comparison_df = comparison_df[columns_to_display]
```

```
comparison_df.set_index('Model', inplace=True)
comparison_df.T
```

```
[22]: Model      Random Forest      SVM      LSTM
      TP          12.000000    13.000000    14.000000
      TN          44.000000    44.000000    35.000000
      FP           6.000000     6.000000    15.000000
      FN          14.000000    13.000000    12.000000
      TPR          0.461538     0.500000     0.538462
      TNR          0.880000     0.880000     0.700000
      FPR          0.120000     0.120000     0.300000
      FNR          0.538462     0.500000     0.461538
      Accuracy     0.736842     0.750000     0.644737
      Precision     0.666667     0.684211     0.482759
      Recall       15.000000    14.000000    13.000000
      F1 Score      0.545455     0.577778     0.509091
      bacc          0.670769     0.690000     0.619231
      tss           0.341538     0.380000     0.238462
      hss           0.368771     0.406250     0.232036
```