

# modelpart1

April 24, 2025

## 0.0.1 Objective:

Build different models and evaluate each one of them using different metrics.

```
[1]: ## import essential libraries

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from functions_model import *
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
```

```
[2]: ## Reading and printing csv file
df = pd.read_csv("FileModel.csv")
df.head()
```

```
[2]:
```

	Gender	Age	Academic Pressure	CGPA	Study Satisfaction	Degree \
0	Male	33	5	8.97	2	B.Pharm
1	Female	24	2	5.90	5	BSc
2	Male	31	3	7.03	5	BA
3	Female	28	3	5.59	2	BCA
4	Female	25	4	8.13	3	M.Tech

	Have you ever had suicidal thoughts ?	Work/Study Hours	Financial Stress \
0	Yes	3	1
1	No	3	2
2	No	9	1
3	Yes	4	5
4	Yes	1	1

	Family History of Mental Illness	State	Sleep Duration_encoded \
0	No	Andhra Pradesh	1
1	Yes	Karnataka	1
2	Yes	Jammu and Kashmir	0
3	Yes	Uttar Pradesh	3
4	No	Rajasthan	1

	Dietary Habits_encoded	Depression
0	2	1
1	1	0
2	2	0
3	1	1
4	1	0

```
[3]: print(df.dtypes)
```

```
Gender                object
Age                  int64
Academic Pressure      int64
CGPA                 float64
Study Satisfaction    int64
Degree              object
Have you ever had suicidal thoughts ? object
Work/Study Hours      int64
Financial Stress       int64
Family History of Mental Illness object
State                object
Sleep Duration_encoded int64
Dietary Habits_encoded int64
Depression            int64
dtype: object
```

## 0.0.2 One Hot Encoding Columns

Columns need to be one hot encoded.

- Gender
- Degree
- Have you ever had suicidal thoughts
- Family History of Mental Illness
- State

```
[4]: df = one_hot_encode_column(df, 'Gender')
df = one_hot_encode_column(df, 'Degree')
df = one_hot_encode_column(df, 'Have you ever had suicidal thoughts ?')
df = one_hot_encode_column(df, 'Family History of Mental Illness')
df = one_hot_encode_column(df, 'State')

df.head()
```

```
[4]:   Age  Academic Pressure  CGPA  Study Satisfaction  Work/Study Hours  \
0   33                5  8.97                2            3
1   24                2  5.90                5            3
2   31                3  7.03                5            9
3   28                3  5.59                2            4
```

```

4    25                4    8.13                3                1

    Financial Stress  Sleep Duration_encoded  Dietary Habits_encoded  \
0                1                1                2
1                2                1                1
2                1                0                2
3                5                3                1
4                1                1                1

    Depression  Gender_Female  ...  State_Jammu and Kashmir  State_Karnataka  \
0                1                0  ...                0                0
1                0                1  ...                0                1
2                0                0  ...                1                0
3                1                1  ...                0                0
4                0                1  ...                0                0

    State_Madhya Pradesh  State_Maharashtra  State_Punjab  State_Rajasthan  \
0                0                0                0                0
1                0                0                0                0
2                0                0                0                0
3                0                0                0                0
4                0                0                0                1

    State_Tamil Nadu  State_Telangana  State_Uttar Pradesh  State_West Bengal
0                0                0                0                0
1                0                0                0                0
2                0                0                0                0
3                0                0                1                0
4                0                0                0                0

```

[5 rows x 57 columns]

```

[ ]: ## checking all the datatypes after one hot encoding
      print(df.dtypes)

```

```

Age                int64
Academic Pressure   int64
CGPA                float64
Study Satisfaction  int64
Work/Study Hours    int64
Financial Stress     int64
Sleep Duration_encoded int64
Dietary Habits_encoded int64
Depression          int64
Gender_Female       int64
Gender_Male         int64
Degree_'Class 12'   int64
Degree_B.Arch       int64

```

Degree_B.Com	int64
Degree_B.Ed	int64
Degree_B.Pharm	int64
Degree_B.Tech	int64
Degree_BA	int64
Degree_BBA	int64
Degree_BCA	int64
Degree_BE	int64
Degree_BHM	int64
Degree_BSc	int64
Degree_LLB	int64
Degree_LLM	int64
Degree_M.Com	int64
Degree_M.Ed	int64
Degree_M.Pharm	int64
Degree_M.Tech	int64
Degree_MA	int64
Degree_MBA	int64
Degree_MBBS	int64
Degree_MCA	int64
Degree_MD	int64
Degree_ME	int64
Degree_MHM	int64
Degree_MSc	int64
Degree_PhD	int64
Have you ever had suicidal thoughts ?_No	int64
Have you ever had suicidal thoughts ?_Yes	int64
Family History of Mental Illness_No	int64
Family History of Mental Illness_Yes	int64
State_Andhra Pradesh	int64
State_Bihar	int64
State_Delhi	int64
State_Gujarat	int64
State_Haryana	int64
State_Jammu and Kashmir	int64
State_Karnataka	int64
State_Madhya Pradesh	int64
State_Maharashtra	int64
State_Punjab	int64
State_Rajasthan	int64
State_Tamil Nadu	int64
State_Telangana	int64
State_Uttar Pradesh	int64
State_West Bengal	int64
dtype: object	

### 0.0.3 Models to be used:

- Logistic Regression
- Decision Tree
- Random Forest
- SVM

### 0.0.4 Baseline Model

Constructing a baseline model, using Logistic Regression because we have binary target variable.

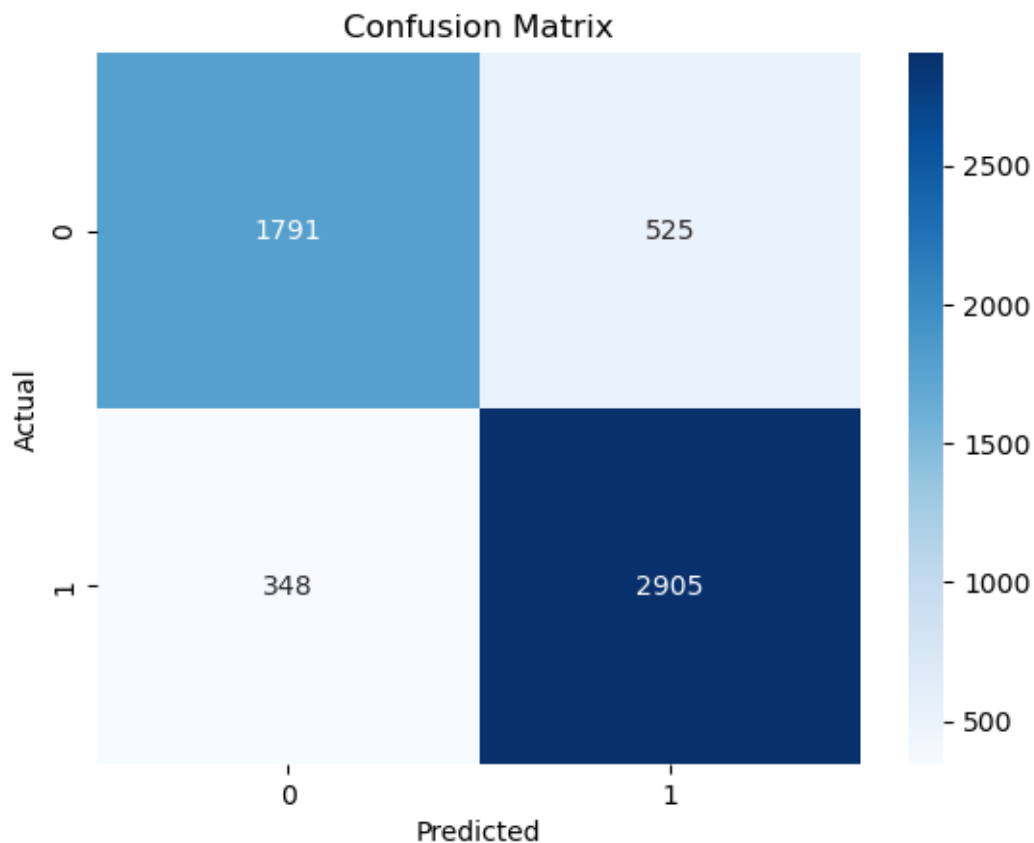
### 0.0.5 Logistic Regression

```
[6]: X = df.drop("Depression", axis=1)
y = df["Depression"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳ random_state= 42)
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
model1 = LogisticRegression()

model_evaulate(model1, X_train_scaled, y_train, X_test_scaled, y_test)
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.84	0.77	0.80	2316
1	0.85	0.89	0.87	3253
accuracy			0.84	5569
macro avg	0.84	0.83	0.84	5569
weighted avg	0.84	0.84	0.84	5569



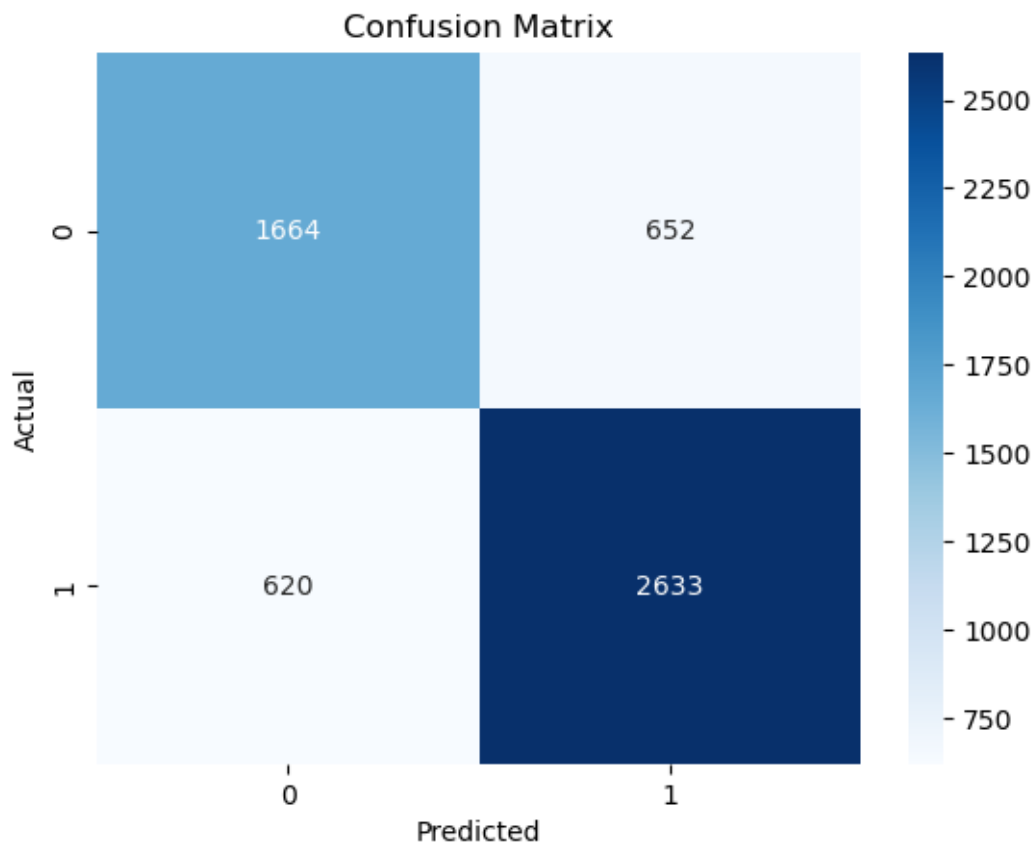
### 0.0.6 Decision Tree

```
[7]: from sklearn.tree import DecisionTreeClassifier
```

```
[8]: model2 = DecisionTreeClassifier(criterion= 'gini', random_state=42)
      model_evaulate(model2, X_train, y_train, X_test, y_test)
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.73	0.72	0.72	2316
1	0.80	0.81	0.81	3253
accuracy			0.77	5569
macro avg	0.77	0.76	0.76	5569
weighted avg	0.77	0.77	0.77	5569

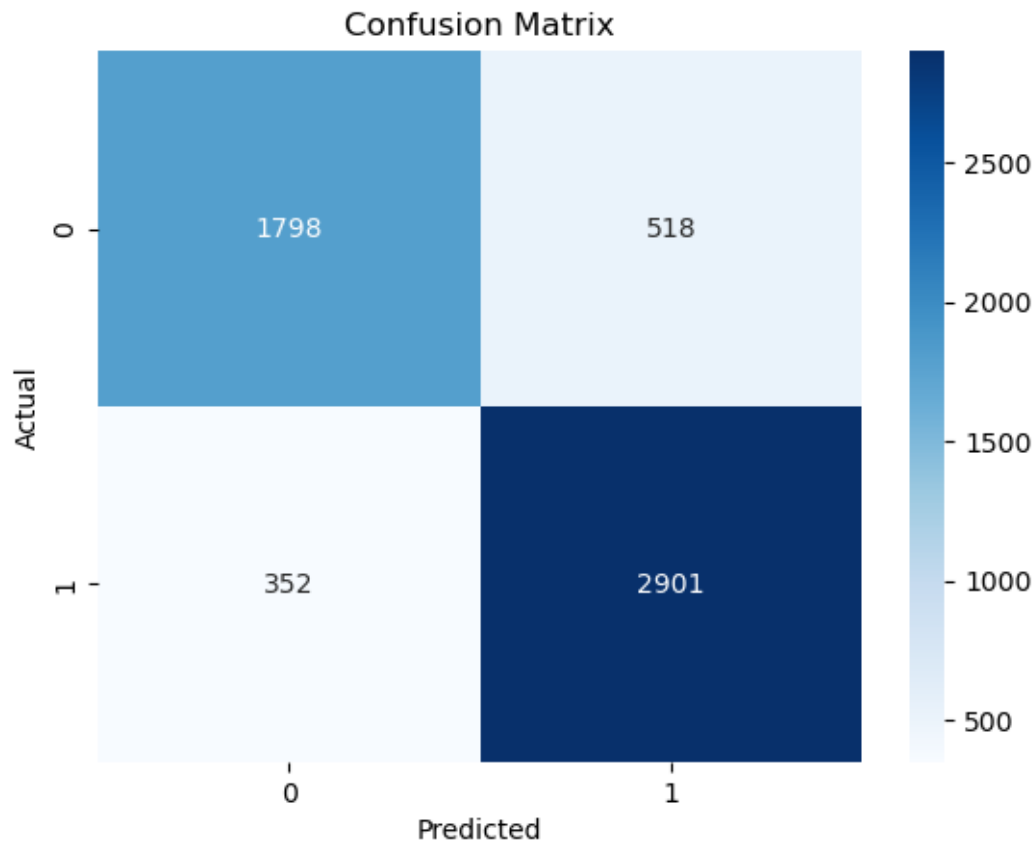


### 0.0.7 Random Forest

```
[9]: from sklearn.ensemble import RandomForestClassifier
model3 = RandomForestClassifier(random_state=42)
model_evaulate(model3, X_train, y_train, X_test, y_test)
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.84	0.78	0.81	2316
1	0.85	0.89	0.87	3253
accuracy			0.84	5569
macro avg	0.84	0.83	0.84	5569
weighted avg	0.84	0.84	0.84	5569



### 0.0.8 Hyperparameter Tuning for Random Forest

```
[ ]: RFC = RandomForestClassifier()

params = {
    'n_estimators': [100, 300],
    'max_depth': [None, 20, 30],
    'min_samples_split': [2, 5],
    'min_samples_leaf': [1, 4],
    'max_features': ['sqrt', 0.8],
    'bootstrap': [True],
    'class_weight': ['balanced'],
    'criterion': ['gini', 'entropy']
}

## Using GridSearch
grid_search = GridSearchCV(
    estimator=RFC,
    param_grid=params,
    cv=5,
```



```

        scoring='recall',
        n_jobs=-1,
        verbose=2
    )

```

Finding the best model using `best_estimator_` function

- Note: Uncomment the line below to see the best hyperparameters

```
[12]: ##model_evaluate(grid_search, X_train, y_train, X_test, y_test)
```

### 0.0.9 Cross Validation:

Cross Validating the data. Trying to improve recall.

```

[ ]: from sklearn.model_selection import StratifiedKFold, cross_val_score
     cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=42)
     RF = RandomForestClassifier(class_weight='balanced', max_depth=30,
     ↪ n_estimators=300)
     best_RF_cv = cross_val_score(RF, X_train, y_train, cv=cv, scoring='recall')
     ↪      ## using recall
     print(f"Fold Metrics:\n{best_RF_cv}\n")
     print(f"Average Score:\n{best_RF_cv.mean()}\n")
     ↪      ## Average Score of Recall

```

Fold Metrics:

```
[0.88995399 0.88113497 0.89033742 0.87423313 0.88190184]
```

Average Score:

```
0.8835122699386504
```

```

[ ]: ## Printing each iteration, since the dataset was split into 5 parts
     crossval_evaluate(RF, X, y)

```

Fold 1 Recall: 0.8776

Confusion Matrix:

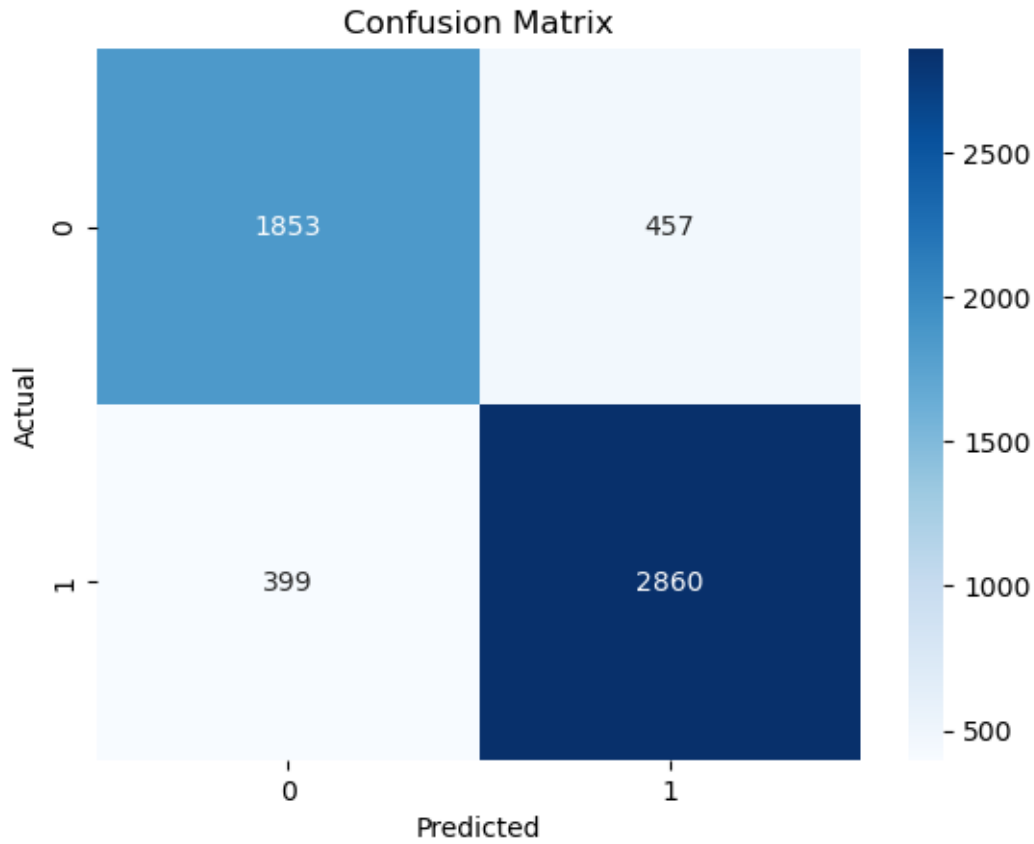
```
[[1853  457]
```

```
 [ 399 2860]]
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.82	0.80	0.81	2310
1	0.86	0.88	0.87	3259
accuracy			0.85	5569
macro avg	0.84	0.84	0.84	5569

weighted avg	0.85	0.85	0.85	5569
--------------	------	------	------	------



Fold 2 Recall: 0.8825

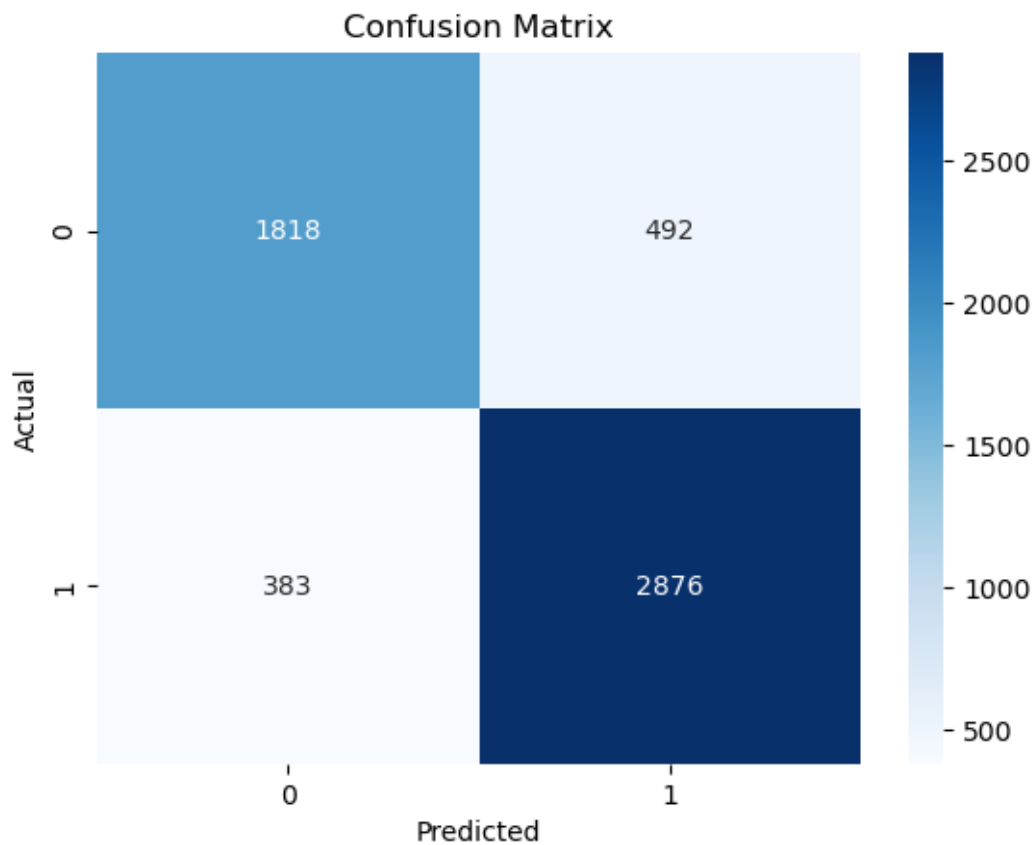
Confusion Matrix:

```
[[1818  492]
```

```
 [ 383 2876]]
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.83	0.79	0.81	2310
1	0.85	0.88	0.87	3259
accuracy			0.84	5569
macro avg	0.84	0.83	0.84	5569
weighted avg	0.84	0.84	0.84	5569



Fold 3 Recall: 0.8699

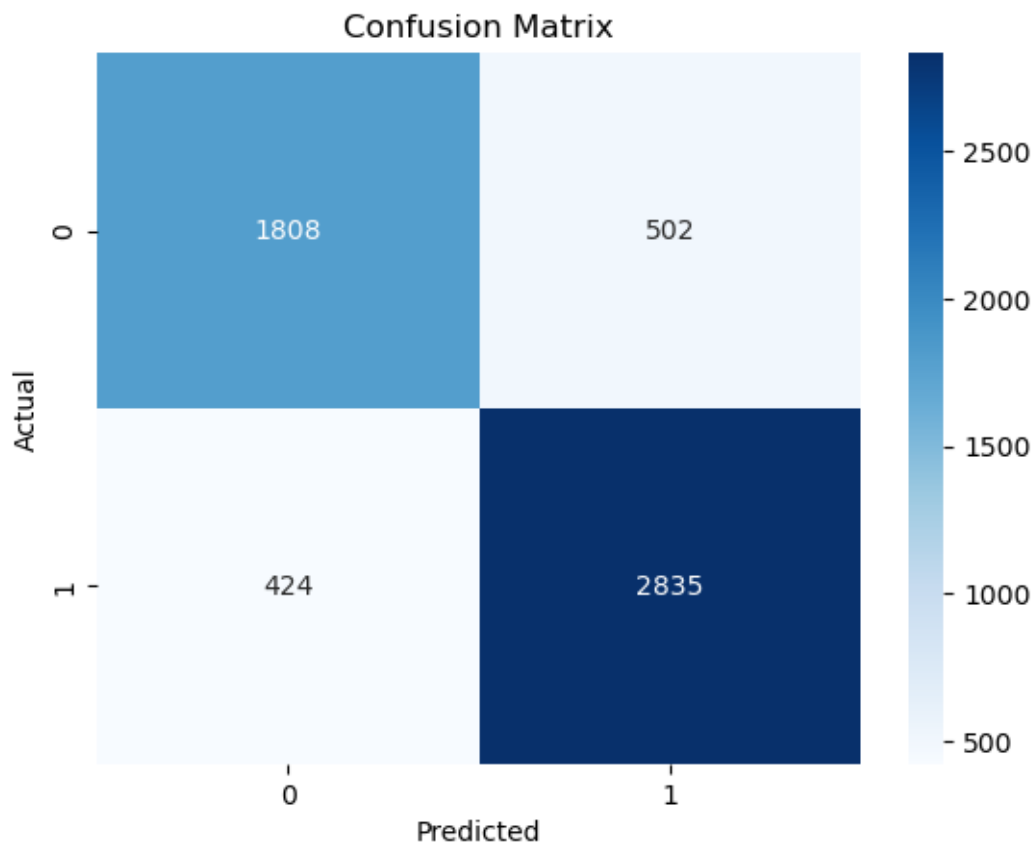
Confusion Matrix:

```
[[1808  502]
```

```
 [ 424 2835]]
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.81	0.78	0.80	2310
1	0.85	0.87	0.86	3259
accuracy			0.83	5569
macro avg	0.83	0.83	0.83	5569
weighted avg	0.83	0.83	0.83	5569



Fold 4 Recall: 0.8864

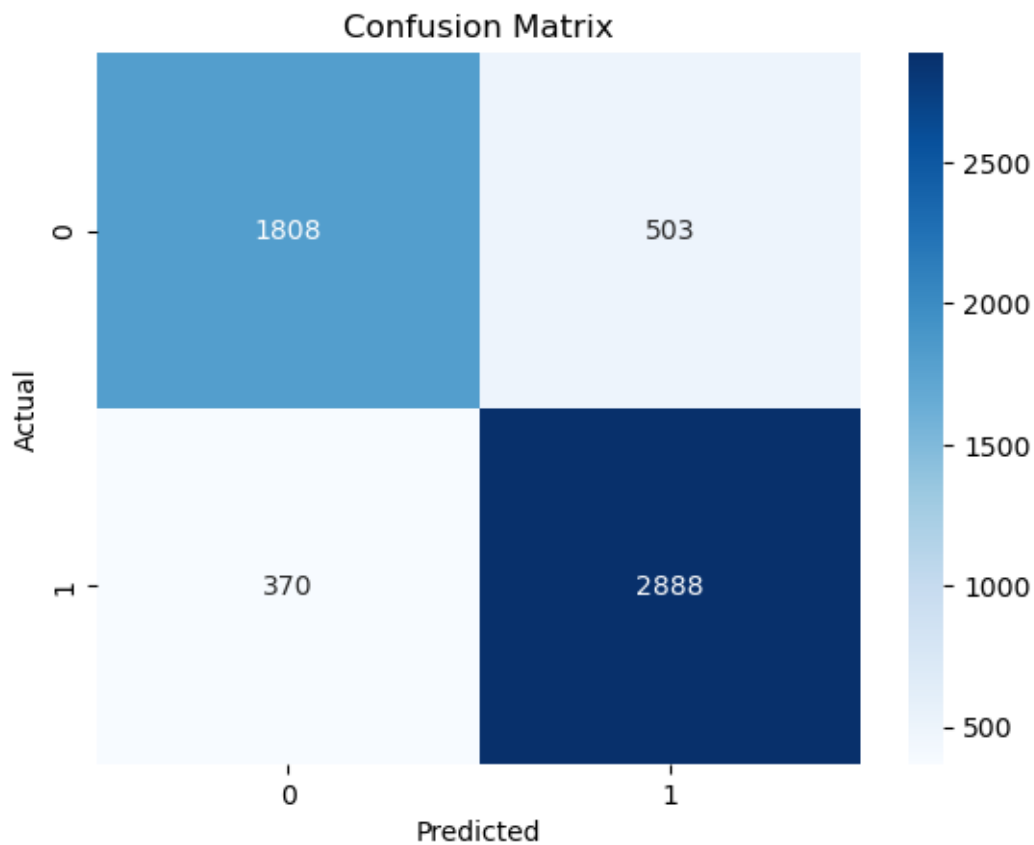
Confusion Matrix:

```
[[1808  503]
```

```
 [ 370 2888]]
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.83	0.78	0.81	2311
1	0.85	0.89	0.87	3258
accuracy			0.84	5569
macro avg	0.84	0.83	0.84	5569
weighted avg	0.84	0.84	0.84	5569



Fold 5 Recall: 0.8849

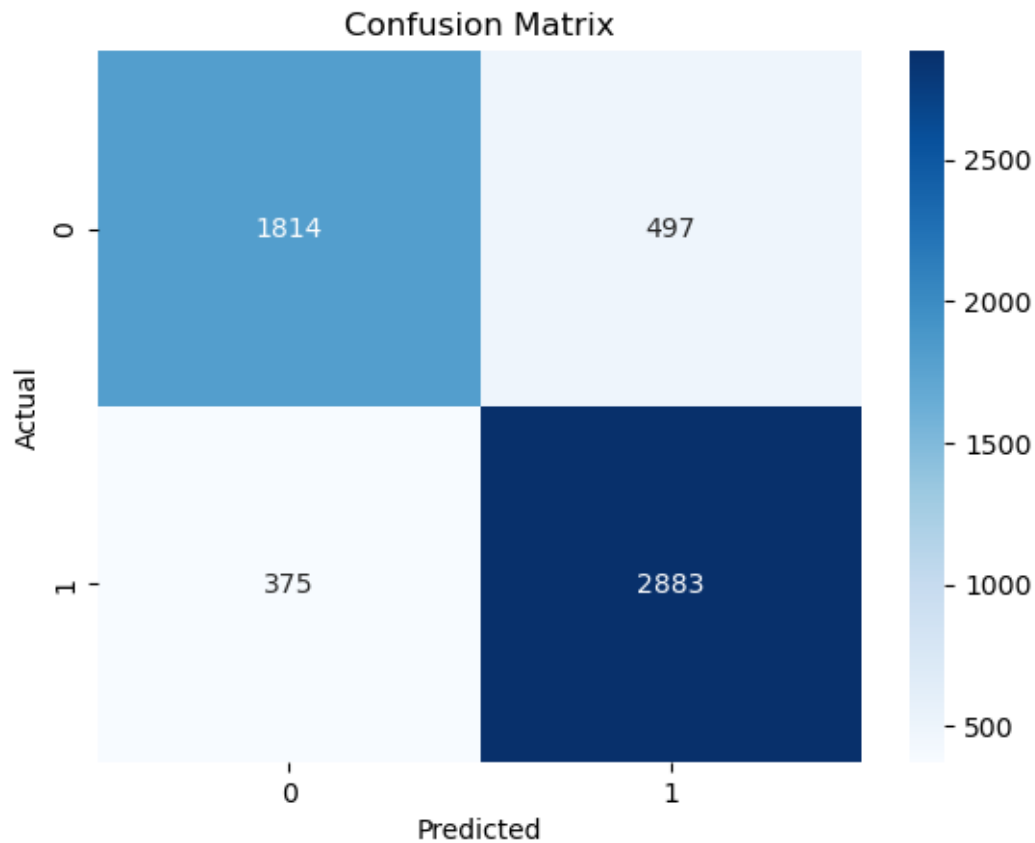
Confusion Matrix:

```
[[1814  497]
```

```
 [ 375 2883]]
```

<IPython.core.display.HTML object>

	precision	recall	f1-score	support
0	0.83	0.78	0.81	2311
1	0.85	0.88	0.87	3258
accuracy			0.84	5569
macro avg	0.84	0.83	0.84	5569
weighted avg	0.84	0.84	0.84	5569



Cross-Validation Summary  
All Fold Recalls: [0.8775698066891685, 0.8824792881251918, 0.869898741945382,  
0.8864333947206875, 0.8848987108655617]  
Mean Recall: 0.8803  
Best Fold Recall: 0.8864  
Best Confusion Matrix:  
[[1808 503]  
 [ 370 2888]]