# Credit Scoring Model – Give Me Some Credit Dataset

```
# Libraries
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, cross_val_score
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report, confusion_matrix, roc_auc_score, roc_curve

# Ignore warnings
import warnings
warnings.filterwarnings("ignore")
```

# 1. Load Dataset

```
import kagglehub
# Download dataset
path = kagglehub.dataset_download("saleha07/give-me-some-credit")
print("<a href="Path">Path</a> to dataset: ", path)
# Load data
df = pd.read_csv(f"{path}/cs-training.csv")
df.head()
Downloading from https://www.kaggle.com/api/v1/datasets/download/saleha07/give-me-some-credit?dataset_version_number=1....
             5.16M/5.16M [00:00<00:00, 5.43MB/s]Extracting files...
☑ Path to dataset: /root/.cache/kagglehub/datasets/saleha07/give-me-some-credit/versions/1
             SeriousDlqin2yrs RevolvingUtilizationOfUnsecuredLines age Nummber 01 1 1 me 20 59DaysPastDueNotWorse
                                                                                                      DebtRatio MonthlyIncome Numb
0
           1
                              1
                                                               0.766127
                                                                         45
                                                                                                       0.802982
                                                                                                                         9120.0
           2
                                                                                                                         2600.0
                              0
                                                               0.957151
                                                                         40
                                                                                                       0.121876
 1
 2
           3
                              0
                                                               0.658180
                                                                         38
                                                                                                        0.085113
                                                                                                                         3042.0
 3
           4
                              0
                                                                                                       0.036050
                                                                                                                         3300.0
                                                               0.233810
                                                                         30
           5
                                                               0.907239 49
                                                                                                       0.024926
                                                                                                                        63588.0
```

#### Explore Dataset

```
# Shape and info
print("Shape:", df.shape)
print(df.info())

# Target distribution
df['SeriousDlqin2yrs'].value_counts(normalize=True).plot(kind='bar', title="Target Distribution")
plt.show()
```

```
Shape: (150000, 12)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150000 entries, 0 to 149999
Data columns (total 12 columns):
                                          Non-Null Count
   Column
                                                           Dtype
#
0
    Unnamed: 0
                                          150000 non-null int64
    SeriousDlqin2yrs
                                          150000 non-null
                                                           int64
    RevolvingUtilizationOfUnsecuredLines
                                          150000 non-null
                                                           float64
3
                                          150000 non-null int64
    NumberOfTime30-59DaysPastDueNotWorse
                                          150000 non-null int64
                                          150000 non-null float64
    DebtRatio
6
    MonthlyIncome
                                          120269 non-null
                                                           float64
    NumberOfOpenCreditLinesAndLoans
                                          150000 non-null int64
    NumberOfTimes90DaysLate
                                          150000 non-null
8
                                                           int64
    NumberRealEstateLoansOrLines
                                          150000 non-null int64
10
    NumberOfTime60-89DaysPastDueNotWorse
                                          150000 non-null
                                                           int64
11 NumberOfDependents
                                          146076 non-null float64
dtypes: float64(4), int64(8)
memory usage: 13.7 MB
                           Target Distribution
 0.8
 0.6
 0.4
 0.2
 0.0
                              SeriousDlqin2yrs
```

## Data Cleaning

```
# Drop ID column if present
if 'Unnamed: 0' in df.columns:
    df.drop(columns=['Unnamed: 0'], inplace=True)
# Check missing values
print(df.isnull().sum())
# Impute missing values
df['MonthlyIncome'].fillna(df['MonthlyIncome'].median(), inplace=True)
df['NumberOfDependents'].fillna(df['NumberOfDependents'].median(), inplace=True)
SeriousDlgin2yrs
                                             0
RevolvingUtilizationOfUnsecuredLines
                                             0
age
                                             0
NumberOfTime30-59DaysPastDueNotWorse
                                             a
DebtRatio
                                             а
MonthlyIncome
                                        29731
NumberOfOpenCreditLinesAndLoans
                                             0
NumberOfTimes90DaysLate
                                             0
NumberRealEstateLoansOrLines
NumberOfTime60-89DaysPastDueNotWorse
                                          3924
NumberOfDependents
dtype: int64
```

### Feature Engineering

```
# Create new features
df['IncomePerDependent'] = df['MonthlyIncome'] / (df['NumberOfDependents'] + 1)
df['DebtToIncomeRatio'] = df['DebtRatio'] * df['MonthlyIncome']
# Cap unrealistic ages
```

```
df = df[df['age'] >= 18]
```

### Train-Test Split

### Baseline Models

#### Logistic Regression

```
log_reg = LogisticRegression(class_weight='balanced', max_iter=1000)
log_reg.fit(X_train, y_train)
y_pred_lr = log_reg.predict(X_test)
y_prob_lr = log_reg.predict_proba(X_test)[:,1]
print("♦ Logistic Regression Report:")
print(classification_report(y_test, y_pred_lr))
print("ROC-AUC:", roc_auc_score(y_test, y_prob_lr))
♦ Logistic Regression Report:
             precision
                        recall f1-score support
                  0.97
                          0.78
                                              27995
          0
                                     0.86
                        0.67
                  0.18
                                     0.28
                                              2005
          1
                                              30000
                                     0.77
   accuracy
                  0.57
                           9.72
                                              30000
  macro avg
                                     0.57
weighted avg
                  0.92
                           0.77
                                     0.82
                                              30000
ROC-AUC: 0.8005925799895688
```

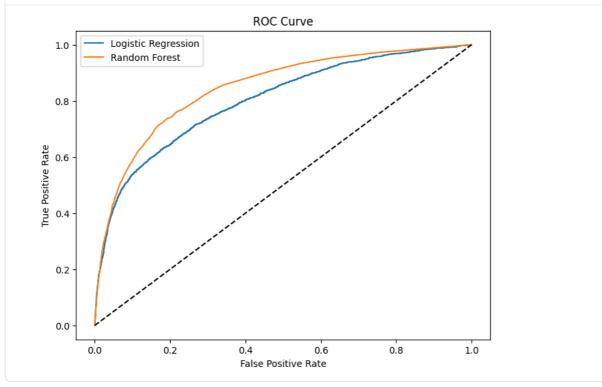
#### Random Forest

```
\verb|rf = RandomForestClassifier(n_estimators=200, class_weight='balanced', random_state=42)| \\
rf.fit(X_train, y_train)
y_pred_rf = rf.predict(X_test)
y_prob_rf = rf.predict_proba(X_test)[:,1]
print("♠ Random Forest Report:")
print(classification_report(y_test, y_pred_rf))
print("ROC-AUC:", roc_auc_score(y_test, y_prob_rf))
♦ Random Forest Report:
             precision
                         recall f1-score support
           0
                  0.94
                            0.99
                                      0.97
                                               27995
                  0.55
                            0.15
                                      0.24
                                                2005
                                       0.94
                                                30000
   accuracy
                  0.75
                            0.57
                                       0.60
                                                30000
   macro avg
                                               30000
weighted avg
                 0.92
                            0.94
                                      0.92
ROC-AUC: 0.8456342088162341
```

\* ROC Curve

```
fpr_lr, tpr_lr, _ = roc_curve(y_test, y_prob_lr)
fpr_rf, tpr_rf, _ = roc_curve(y_test, y_prob_rf)

plt.figure(figsize=(8,6))
plt.plot(fpr_lr, tpr_lr, label="Logistic Regression")
plt.plot(fpr_rf, tpr_rf, label="Random Forest")
plt.plot([0,1],[0,1],'k--')
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ROC Curve")
plt.legend()
plt.show()
```



```
import joblib

# Save model
joblib.dump(rf, "Credit Scoring Model.pkl")

# Load model later
loaded_rf = joblib.load("random_forest_model.pkl")

# Test loaded model
print("ROC-AUC (loaded model):", roc_auc_score(y_test, loaded_rf.predict_proba(X_test)[:,1]))
ROC-AUC (loaded model): 0.8456342088162341
```