

## Load the dataset

```
In [1]: import pandas as pd
```

```
In [2]: # Load the dataset
data = "creditcard.csv"
data = pd.read_csv(data)
```

```
In [3]: # Display basic information about the dataset
print(data.info())

# Display the first few rows of the dataset
print(data.head())
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 284807 entries, 0 to 284806
```

```
Data columns (total 31 columns):
```

```
#    Column    Non-Null Count  Dtype
```

```
---  -
0    Time      284807 non-null  float64
1    V1        284807 non-null  float64
2    V2        284807 non-null  float64
3    V3        284807 non-null  float64
4    V4        284807 non-null  float64
5    V5        284807 non-null  float64
6    V6        284807 non-null  float64
7    V7        284807 non-null  float64
8    V8        284807 non-null  float64
9    V9        284807 non-null  float64
10   V10       284807 non-null  float64
11   V11       284807 non-null  float64
12   V12       284807 non-null  float64
13   V13       284807 non-null  float64
14   V14       284807 non-null  float64
15   V15       284807 non-null  float64
16   V16       284807 non-null  float64
17   V17       284807 non-null  float64
18   V18       284807 non-null  float64
19   V19       284807 non-null  float64
20   V20       284807 non-null  float64
21   V21       284807 non-null  float64
22   V22       284807 non-null  float64
23   V23       284807 non-null  float64
24   V24       284807 non-null  float64
25   V25       284807 non-null  float64
26   V26       284807 non-null  float64
27   V27       284807 non-null  float64
28   V28       284807 non-null  float64
29   Amount    284807 non-null  float64
30   Class     284807 non-null  int64
```

```
dtypes: float64(30), int64(1)
```

```
memory usage: 67.4 MB
```

```
None
```

```
Time      V1      V2      V3      V4      V5      V6      V7  \
0    0.0 -1.359807 -0.072781  2.536347  1.378155 -0.338321  0.462388  0.239599
```

1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941

  

	V8	V9	...	V21	V22	V23	V24	V25	\
0	0.098698	0.363787	...	-0.018307	0.277838	-0.110474	0.066928	0.128539	
1	0.085102	-0.255425	...	-0.225775	-0.638672	0.101288	-0.339846	0.167170	
2	0.247676	-1.514654	...	0.247998	0.771679	0.909412	-0.689281	-0.327642	
3	0.377436	-1.387024	...	-0.108300	0.005274	-0.190321	-1.175575	0.647376	
4	-0.270533	0.817739	...	-0.009431	0.798278	-0.137458	0.141267	-0.206010	

  

	V26	V27	V28	Amount	Class
0	-0.189115	0.133558	-0.021053	149.62	0
1	0.125895	-0.008983	0.014724	2.69	0
2	-0.139097	-0.055353	-0.059752	378.66	0
3	-0.221929	0.062723	0.061458	123.50	0
4	0.502292	0.219422	0.215153	69.99	0

[5 rows x 31 columns]

## Data Preprocessing

```
In [4]: from sklearn.preprocessing import StandardScaler

# Standardize the 'Amount' column
scaler = StandardScaler()
data['Amount'] = scaler.fit_transform(data['Amount'].values.reshape(-1, 1))

# Drop unnecessary columns (if any)
# For example, if 'Time' is not relevant, you can drop it
data = data.drop(['Time'], axis=1)
```

```
In [5]: pip install imbalanced-learn
```

```
Requirement already satisfied: imbalanced-learn in c:\users\h s computer s hyd\anaconda3\lib\site-packages (0.11.0)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\h s computer s hyd\anaco
nda3\lib\site-packages (from imbalanced-learn) (2.2.0)
Requirement already satisfied: joblib>=1.1.1 in c:\users\h s computer s hyd\anaconda3\li
b\site-packages (from imbalanced-learn) (1.3.2)
Requirement already satisfied: scikit-learn>=1.0.2 in c:\users\h s computer s hyd\anacon
da3\lib\site-packages (from imbalanced-learn) (1.3.2)
Requirement already satisfied: scipy>=1.5.0 in c:\users\h s computer s hyd\anaconda3\lib
\site-packages (from imbalanced-learn) (1.7.1)
Requirement already satisfied: numpy>=1.17.3 in c:\users\h s computer s hyd\anaconda3\li
b\site-packages (from imbalanced-learn) (1.22.4)
Note: you may need to restart the kernel to use updated packages.
```

## Handle Class Imbalance

```
In [6]: # Count the number of fraudulent and genuine transactions
fraud_count = data['Class'].sum()
genuine_count = len(data) - fraud_count

# Display class distribution
```

```
print(f"Fraudulent transactions: {fraud_count}")
print(f"Genuine transactions: {genuine_count}")
```

Fraudulent transactions: 492  
Genuine transactions: 284315

## Split the dataset

```
In [7]: from sklearn.model_selection import train_test_split

# Separate features (X) and target variable (y)
X = data.drop('Class', axis=1)
y = data['Class']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4
```

```
In [8]: from sklearn.utils import resample

# Combine X_train and y_train for resampling
combined_data = pd.concat([X_train, y_train], axis=1)

# Separate majority and minority classes
majority_class = combined_data[combined_data['Class'] == 0]
minority_class = combined_data[combined_data['Class'] == 1]

# Upsample the minority class
minority_upsampled = resample(minority_class, replace=True, n_samples=len(majority_class))

# Combine the upsampled minority class with the majority class
upsampled_data = pd.concat([majority_class, minority_upsampled])

# Separate features (X_resampled) and target variable (y_resampled)
X_resampled = upsampled_data.drop('Class', axis=1)
y_resampled = upsampled_data['Class']

# Check the class distribution after upsampling
print("Class distribution after upsampling:")
print(y_resampled.value_counts())
```

Class distribution after upsampling:  
0 227451  
1 227451  
Name: Class, dtype: int64

## Build a Random Forest Classifier

```
In [9]: from sklearn.ensemble import RandomForestClassifier

# Initialize and train the model
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_resampled, y_resampled)
```

```
Out[9]: ▼    RandomForestClassifier
RandomForestClassifier(random_state=42)
```

## Evaluate the model

```
In [10]: from sklearn.metrics import classification_report, confusion_matrix

# Make predictions on the test set
y_pred = model.predict(X_test)

# Display classification report and confusion matrix
print("Classification Report:")
print(classification_report(y_test, y_pred))

print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))
```

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	56864
1	0.99	0.80	0.88	98
accuracy			1.00	56962
macro avg	0.99	0.90	0.94	56962
weighted avg	1.00	1.00	1.00	56962

Confusion Matrix:

```
[[56863  1]
 [  20  78]]
```