


▼ Import Librarys

```
import numpy as np
import tensorflow as tf
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

▼ Import Data

```
data=pd.read_csv('/content/Churn_Modelling.csv')
data.head()
```



| | RowNumber | CustomerId | Surname | CreditScore | Geography | Gender | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiveMember | EstimatedSalary |
|---|-----------|------------|----------|-------------|-----------|--------|-----|--------|-----------|---------------|-----------|----------------|-----------------|
| 0 | 1 | 15634602 | Hargrave | 619 | France | Female | 42 | 2 | 0.00 | 1 | 1 | 1 | |
| 1 | 2 | 15647311 | Hill | 608 | Spain | Female | 41 | 1 | 83807.86 | 1 | 0 | 1 | |
| 2 | 3 | 15619304 | Onio | 502 | France | Female | 42 | 8 | 159660.80 | 3 | 1 | 0 | |
| 3 | 4 | 15701354 | Boni | 699 | France | Female | 39 | 1 | 0.00 | 2 | 0 | 0 | |
| 4 | 5 | 15737888 | Mitchell | 850 | Spain | Female | 43 | 2 | 125510.82 | 1 | 1 | 1 | |

▼ Dropping unimportant features

```
data=data.drop(columns=['RowNumber','CustomerId','Surname'])
data.head()
```

| | CreditScore | Geography | Gender | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiveMember | EstimatedSalary | Exited |
|---|-------------|-----------|--------|-----|--------|-----------|---------------|-----------|----------------|-----------------|--------|
| 0 | 619 | France | Female | 42 | 2 | 0.00 | 1 | 1 | 1 | 101348.88 | 1 |
| 1 | 608 | Spain | Female | 41 | 1 | 83807.86 | 1 | 0 | 1 | 112542.58 | 0 |
| 2 | 502 | France | Female | 42 | 8 | 159660.80 | 3 | 1 | 0 | 113931.57 | 1 |
| 3 | 699 | France | Female | 39 | 1 | 0.00 | 2 | 0 | 0 | 93826.63 | 0 |
| 4 | 850 | Spain | Female | 43 | 2 | 125510.82 | 1 | 1 | 1 | 79084.10 | 0 |

▼ One Hot Encoding

```
data['Gender']=data['Gender'].apply(lambda x : 0 if x=='Female' else 1)
data['Gender']=data['Gender'].astype(int)
data.head()
```

| | CreditScore | Geography | Gender | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiveMember | EstimatedSalary | Exited |
|---|-------------|-----------|--------|-----|--------|-----------|---------------|-----------|----------------|-----------------|--------|
| 0 | 619 | France | 0 | 42 | 2 | 0.00 | 1 | 1 | 1 | 101348.88 | 1 |
| 1 | 608 | Spain | 0 | 41 | 1 | 83807.86 | 1 | 0 | 1 | 112542.58 | 0 |
| 2 | 502 | France | 0 | 42 | 8 | 159660.80 | 3 | 1 | 0 | 113931.57 | 1 |
| 3 | 699 | France | 0 | 39 | 1 | 0.00 | 2 | 0 | 0 | 93826.63 | 0 |
| 4 | 850 | Spain | 0 | 43 | 2 | 125510.82 | 1 | 1 | 1 | 79084.10 | 0 |

```
from sklearn.preprocessing import LabelEncoder
```

```
label_encoder = LabelEncoder()
data['Geography']=label_encoder.fit_transform(data['Geography'])
data.head()
```

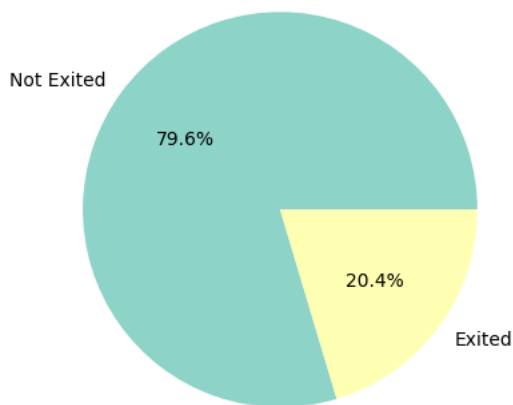
| | CreditScore | Geography | Gender | Age | Tenure | Balance | NumOfProducts | HasCrCard | IsActiveMember | EstimatedSalary | Exited |
|---|-------------|-----------|--------|-----|--------|-----------|---------------|-----------|----------------|-----------------|--------|
| 0 | 619 | 0 | 0 | 42 | 2 | 0.00 | 1 | 1 | 1 | 101348.88 | 1 |
| 1 | 608 | 2 | 0 | 41 | 1 | 83807.86 | 1 | 0 | 1 | 112542.58 | 0 |
| 2 | 502 | 0 | 0 | 42 | 8 | 159660.80 | 3 | 1 | 0 | 113931.57 | 1 |
| 3 | 699 | 0 | 0 | 39 | 1 | 0.00 | 2 | 0 | 0 | 93826.63 | 0 |
| 4 | 850 | 2 | 0 | 42 | 2 | 125510.82 | 1 | 1 | 1 | 70004.10 | 0 |

▼ Data Analysis

```
value_counts=data['Exited'].value_counts()
plt.pie(value_counts, labels=['Not Exited', 'Exited'], autopct='%1.1f%%', colors=sns.color_palette('Set3'))
```

```
value_counts
```

```
0    7963
1    2037
Name: Exited, dtype: int64
```



```
X=data.drop('Exited',axis=1)
y=data['Exited']
```

▼ Balance Data

```
import numpy as np
from collections import Counter
from imblearn.over_sampling import RandomOverSampler

print("Class distribution before oversampling:", Counter(y))

ros = RandomOverSampler(random_state=42)

X, y = ros.fit_resample(X, y)

print("Class distribution after oversampling:", Counter(y))

Class distribution before oversampling: Counter({0: 7963, 1: 2037})
Class distribution after oversampling: Counter({1: 7963, 0: 7963})
```

▼ Standardize Data

```
X=np.array(X)
X=(X-X.mean())/X.std()
```

▼ Splitting data into train and test

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(X,y,test_size=0.20,random_state=30)
```

▼ model evaluation

```
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
```

```
# Random Forests
random_forest_model = RandomForestClassifier()
random_forest_model.fit(x_train, y_train)
random_forest_pred = random_forest_model.predict(x_test)

print("Random Forests:")
print("Accuracy:", accuracy_score(y_test, random_forest_pred))
print("Classification Report:\n", classification_report(y_test, random_forest_pred))
```

```
Random Forests:
Accuracy: 0.9516635279347144
Classification Report:
              precision    recall  f1-score   support

     0           0.98       0.92      0.95       1593
     1           0.93       0.98      0.95       1593

 accuracy                   0.95       3186
  macro avg           0.95       0.95      0.95       3186
 weighted avg          0.95       0.95      0.95       3186
```