

Step 1: Load the data set, Check if any missing vaues and split it into train & test sets.

```
In [1]: import pandas as pd
df = pd.read_csv("Churn-Modelling.csv")
print(df.shape)
print(df.head())
pd.isnull(df).sum()
```

```
(10000, 7)
   CreditScore  Age  Tenure   Balance  HasCrCard   Salary  Exited
0           619   42      2     0.00         1  101348.88       1
1           608   41      1  83807.86         0  112542.58       0
2           502   42      8  159660.80         1  113931.57       1
3           699   39      1     0.00         0   93826.63       0
4           850   43      2  125510.82         1   79084.10       0
```

```
Out[1]: CreditScore    0
Age                  0
Tenure               0
Balance              0
HasCrCard            0
Salary               0
Exited               0
dtype: int64
```

```
In [2]: x = df.drop("Exited", axis=1)
y = df["Exited"]
print(x.head())
print(y.head())
```

```
   CreditScore  Age  Tenure   Balance  HasCrCard   Salary
0           619   42      2     0.00         1  101348.88
1           608   41      1  83807.86         0  112542.58
2           502   42      8  159660.80         1  113931.57
3           699   39      1     0.00         0   93826.63
4           850   43      2  125510.82         1   79084.10
0          1
1          0
2          1
3          0
4          0
Name: Exited, dtype: int64
```

```
In [3]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.33)
print(x_train.shape, y_train.shape)
print(x_test.shape, y_test.shape)
```

```
(6700, 6) (6700,)
(3300, 6) (3300,)
```

Step 2: Build the model

`sklearn.linear_model.LogisticRegression` class can be used to create model object

`fit()` method can be used to train the model

`predict()` method can be used to predict the outputs for unseen data

```
In [4]: from sklearn.linear_model import LogisticRegression

model = LogisticRegression()
model.fit(x_train, y_train)
```

```
Out[4]: LogisticRegression()
```

Step 3: Test the model performance

we can import various metrics(`confusion_matrix`, `accuracy_score`, `recall_score`, `precision_score`, `f1_score`) from `sklearn.metrics`

```
In [10]: from sklearn.metrics import confusion_matrix, accuracy_score, recall_score, pr
y_pred = model.predict(x_test)
print("Confuusion Matrix:")
print(confusion_matrix(y_test, y_pred))
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Recall:", recall_score(y_test, y_pred))
print("Precision:", precision_score(y_test, y_pred))
print("F1-Score:", f1_score(y_test, y_pred))
```

Confuusion Matrix:

```
[[2587   59]
 [ 616   38]]
```

Accuracy: 0.7954545454545454

Recall: 0.0581039755351682

Precision: 0.3917525773195876

F1-Score: 0.10119840213049268

```
In [8]: cm = confusion_matrix(y_test, y_pred)
tn = cm[0, 0]
tp = cm[1, 1]
fp = cm[0, 1]
fn = cm[1, 0]

print("Accuracy: ", (tn + tp)/(tn + tp + fp + fn))
print("Sensitivity: ", tp / (tp + fn))
print("Specificity: ", tn / (tn + fp))
print("Precision: ", tp / (tp + fp))
```

```
Accuracy:  0.7954545454545454
Sensitivity:  0.0581039755351682
Specificity:  0.9777021919879063
Precision:  0.3917525773195876
```