```
import pandas as pd
import numpy as np
```

from google.colab import files

```
uploaded = files.upload()
```

Choose Files train.csv

train.csv(text/csv) - 61194 bytes, last modified: 10/23/2023 - 100% done
 Saving train.csv to train (1).csv

```
df = pd.read_csv('train.csv')
df.head()
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S	11.
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S	

df.duplicated()

```
0
       False
1
       False
2
       False
3
       False
4
       False
886
       False
887
       False
888
       False
889
       False
890
       False
```

Length: 891, dtype: bool

dtype: int64

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):
               Non-Null Count Dtype
# Column
    PassengerId 891 non-null
    Survived 891 non-null int64
1
2
    Pclass
                 891 non-null int64
    Name
                 891 non-null
                                 object
    Sex
                891 non-null object
                                float64
int64
                 714 non-null
5
    Age
              714 non-null int64
891 non-null int64
891 non-null int64
6
    SibSp
    Parch
                 891 non-null object
891 non-null float64
8
    Ticket
9 Fare
                 204 non-null
                                 object
11 Embarked
                 889 non-null
                                 object
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

```
df['Ticket'].unique()[:50]
        array(['A/5 21171', 'PC 17599', 'STON/O2. 3101282', '113803', '373450',
                  ['A/5 21171', 'PC 17599', 'STON/O2 3101282', '113803', '373450', '330877', '17463', '349909', '347742', '237736', 'PP 9549', '113783', 'A/5. 2151', '347082', '350406', '248706', '382652', '244373', '345763', '2649', '239865', '248698', '330923', '113788', '347077', '2631', '19950', '330959', '349216', 'PC 17601', 'PC 17569', '335677', 'C.A. 24579', 'PC 17604', '113789', '2677', 'A./5. 2152', '345764', '2651', '7546', '11668', '349253', 'SC/Paris 2123', '330958', 'S.C./A.4. 23567', '370371', '14311', '2662', '349237', '3101295'], dtype=object)
df1 = df.drop(columns=['Name','Ticket'])
df1.shape
        (891, 10)
round((df1.isnull().sum()/df1.shape[0])*100,2)
        PassengerId
                                 0.00
        Survived
                                0.00
        Pclass
                                0.00
        Sex
                                0.00
                              19.87
        Age
        SibSp
                                0.00
        Parch
                                0.00
        Fare
                                0.00
        Cabin
                               77.10
        Embarked
                                0.22
        dtype: float64
df2 = df1.drop(columns='Cabin')
df2.dropna(subset=['Embarked'], axis=0, inplace=True)
df2.shape
        (889, 9)
df3 = df2.fillna(df2.Age.mean())
df3.isnull().sum()
        PassengerId
                                0
        Survived
                                0
        Pclass
                                0
        Sex
                               0
                               0
        Age
        SibSp
                                0
                                0
        Parch
        Fare
                               0
        Embarked
        dtype: int64
import matplotlib.pyplot as plt
plt.boxplot(df3['Age'], vert=False)
plt.ylabel('Variable')
plt.xlabel('Age')
plt.title('Box Plot')
plt.show()
```

```
Box Plot
mean = df3['Age'].mean()
std = df3['Age'].std()
lower_bound = mean - std*2
upper_bound = mean + std*2
print('Lower Bound :',lower_bound)
print('Upper Bound:',upper_bound)
df4 = df3[(df3['Age'] >= lower bound)
& (df3['Age'] <= upper_bound)]
     Lower Bound: 3.7054001079256587
     Upper Bound: 55.57878528533277
X = df3[['Pclass','Sex','Age','SibSp','Parch','Fare','Embarked']]
Y = df3['Survived']
{\tt from \ sklearn.preprocessing \ import \ MinMaxScaler}
scaler = MinMaxScaler(feature_range=(0,1))
num_col_ = [col for col in X.columns if X [col].dtype != 'object']
x1 = X
x1[num_col_] = scaler.fit_transform(x1[num_col_])
x1.head()
                                                Traceback (most recent call last)
     NameError
     <ipython-input-15-209de419448c> in <cell line: 5>()
           3 scaler = MinMaxScaler(feature_range=(0,1))
     ----> 5 num_col_ = [col for col in X.columns if X [col].dtype != 'object']
           6 \times 1 = X
     NameError: name 'X' is not defined
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from \ sklearn.linear\_model \ import \ LogisticRegression
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import confusion_matrix, classification_report
from google.colab import files
uploaded= files.upload()
     Choose Files | creditcard.csv
     • creditcard.csv(text/csv) - 532469 bytes, last modified: 10/23/2023 - 100% done
     Saving creditcard.csv to creditcard (2).csv
data = pd.read_csv('creditcard.csv')
print(data.info())
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 999 entries, 0 to 998
     Data columns (total 1 columns):
     # Column
     ---
```

dtypes: object(1)

0 Time, "V1", "V2", "V3", "V4", "V5", "V6", "V7", "V8", "V9", "V10", "V11", "V12", "V13", "V14", "V15", "V16", "V17", "V18", "V19", "V20", "V21", "V22", "V

```
memory usage: 7.9+ KB
         None
print(data.columns)
         Index(['Time,"V1","V2","V3","V4","V5","V6","V7","V8","V9","V10","V11","V12","V13","V14","V15","V16","V17","V18","V19","V20","V22","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V21","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V22","V
data ['normAmount'] = StandardScaler().fit_transform(np.array(data['Amount']).reshape(-1, 1))
data = data.drop(['Time', 'Amount'], axis = 1)
data['Class'].value_counts()
                                                                                       Traceback (most recent call last)
         /usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
              3801
                                              try:
          -> 3802
                                                    return self._engine.get_loc(casted_key)
                                              except KeyError as err:
              3803
                                                                          4 frames
         pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
         pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
         KevError: 'Amount'
         The above exception was the direct cause of the following exception:
         KeyError
                                                                                       Traceback (most recent call last)
         /usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
               3802
                                                      return self._engine.get_loc(casted_key)
               3803
                                              except KeyError as err:
          -> 3804
                                                   raise KeyError(key) from err
               3805
                                              except TypeError:
               3806
                                                     # If we have a listlike key, check indexing error will raise
from sklearn.model_selection import train_test_split
X = data.drop('Class', axis=1)
y = data['Class']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
print("Number of transactions in X_train dataset: ", X_train.shape[0])
print("Number of transactions in y_train dataset: ", y_train.shape[0])
print("Number of transactions in X_test dataset: ", X_test.shape[0])
print("Number of transactions in y_test dataset: ", y_test.shape[0])
                                                                                       Traceback (most recent call last)
         <ipython-input-23-7de74c1a2160> in <cell line: 3>()
                    1 from sklearn.model_selection import train_test_split
          ----> 3 X = data.drop('Class', axis=1)
                   4 y = data['Class']
                                                                          🗘 5 frames
         /usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in drop(self, labels, errors)
               6932
                                      if mask.any():
                                              if errors != "ignore":
               6933
          -> 6934
                                                     raise KeyError(f"{list(labels[mask])} not found in axis")
               6935
                                              indexer = indexer[~mask]
                                    return self.delete(indexer)
               6936
lr = LogisticRegression()
lr fit(X train v train ravel())
```

```
11 · 1 τ ι (Λ_ ι ι α τ ι ι , γ_ ι ι α τ ι ι ι α ν ε τ ( / /
prediction = lr.predict(X test)
print(classification_report(y_test, prediction))
                                               Traceback (most recent call last)
     <ipython-input-24-5f9627479a59> in <cell line: 3>()
           1 lr = LogisticRegression()
     ----> 3 lr.fit(X_train, y_train.ravel())
           5 prediction = lr.predict(X_test)
     NameError: name 'X_train' is not defined
print("Before OverSampling, counts of label '1' : {}".format(sum(y_train == 1)))
print("Before OverSampling, counts of label '0' : {} \n".format(sum(y_train == 0)))
from imblearn.over_sampling import SMOTE
sm = SMOTE(random_state = 2)
X_train_res, y_train_res = (X_train, y_train.ravel())
print('After OverSampling, the shape of train_X: {}'. format(X_train_res.shape))
print('After OverSampling, the shape of train_y: {} \n'. format(y_train_res.shape))
print("After OverSampling, counts of label '1': {}".format(sum(y_train_res == 1)))
print("After OverSampling, counts of label '0': {}".format(sum(y_train_res == 0)))
lr = LogisticRegression()
lr.fit(X_train, y_train.ravel())
predictions = lr.predict(X_test)
print(classification_report(y_test,predictions))
print("Before OverSampling, counts of label '1': {}".format(sum(y_train == 1)))
print("Before OverSampling, counts of label '0': {} \n".format(sum(y_train == 0)))
from \ imblearn.over\_sampling \ import \ SMOTE
sm = SMOTE(random_state = 2)
X_train_res, y_train_res = (X_train, y_train.ravel())
print('After OverSampling, the shape of train_X: {}'.format(X_train_res.shape))
print('After OverSampling, the shape of train_y: {}\n'.format(X_train_res.shape))
print("After OverSampling, counts of label '1': {}". format(sum(y_train_res == 1)))
print("After OverSampling, counts of label '0': {}". format(sum(y_train_res == 0)))
lr1 = LogisticRegression()
lr1.fit(X_train_res, y_train_res.ravel())
prediction = lr1.predict(X_test)
print(classification_report(y_test, prediction))
print("Before Undersampling, counts of label '1': {}".format(sum(y_train == 1 )))
print("Before Undersampling, counts of label '0': {} \n".format(sum(y_train == 0 )))
from imblearn.under_sampling import NearMiss
nr = NearMiss()
X_train_miss, y_train_miss = (X_train, y_train.ravel())
print('After Undersampling, the shape of train_X: {}'.format(X_train_miss.shape))
print('After Undersampling, the shape of train_X: {} \n'.format(y_train_miss.shape))
print("After Undersampling, counts of label '1': {}".format(sum(y_train_miss == 1)))
print("After \ Undersampling, \ counts \ of \ label \ '0': \ \{\}".format(sum(y\_train\_miss == 0)))
```