```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matnlotlib inline
import sklearn
from sklearn.utils import resample
from sklearn.svm import SVC
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.metrics import confusion_matrix, classification_report, accuracy_score
from sklearn import preprocessing
from sklearn.preprocessing import LabelEncoder
from google.colab import drive
drive.mount('/content/gdrive')
    Mounted at /content/gdrive
datas = pd.read_csv("/content/gdrive/MyDrive/bank-full.csv")
datas.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 45211 entries, 0 to 45210
    Data columns (total 17 columns):
     # Column Non-Null Count Dtype
         -----
                    -----
     0 age
                   45211 non-null int64
     1
         job
                   45211 non-null object
        marital
                   45211 non-null object
     3
         education 45211 non-null object
        default
                   45211 non-null object
     4
                   45211 non-null int64
     5
         balance
                   45211 non-null object
     6
        housing
         loan
                   45211 non-null object
     8
         contact
                   45211 non-null object
     9
         day
                   45211 non-null int64
     10 month
                    45211 non-null object
     11 duration 45211 non-null int64
     12 campaign 45211 non-null int64
     13 pdays
                   45211 non-null int64
     14 previous
                   45211 non-null int64
     15 poutcome 45211 non-null object
                   45211 non-null object
     16 y
     dtypes: int64(7), object(10)
     memory usage: 5.9+ MB
datas.head()
```

age		job	marital	education	default	balance	housing	loan	contact	da
0	58	management	married	tertiary	no	2143	yes	no	unknown	
1	44	technician	single	secondary	no	29	yes	no	unknown	
2	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	
3	47	blue-collar	married	unknown	no	1506	yes	no	unknown	
4	33	unknown	single	unknown	no	1	no	no	unknown	
4										•

```
datas = datas.drop_duplicates()
datas = datas[datas.job != 'unknown']
datas = datas[datas.education != 'unknown']
datas.info()
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 43193 entries, 0 to 45210
    Data columns (total 17 columns):
     # Column Non-Null Count Dtype
     0 age
                   43193 non-null int64
     1
         job
                    43193 non-null object
        marital
                   43193 non-null object
         education 43193 non-null object
        default 43193 non-null object
```

43193 non-null int64

balance

```
housing
                    43193 non-null object
         loan
                    43193 non-null object
     8
         contact
                    43193 non-null object
                    43193 non-null int64
         day
     10 month
                    43193 non-null object
     11 duration
                    43193 non-null int64
                    43193 non-null int64
     12 campaign
                    43193 non-null int64
     13 pdays
     14 previous
                    43193 non-null int64
     15 poutcome
                    43193 non-null object
     16 y
                    43193 non-null object
     dtypes: int64(7), object(10)
    memory usage: 5.9+ MB
encoder = preprocessing.LabelEncoder()
datas['job'] = encoder.fit_transform(datas['job'])
datas['marital'] = encoder.fit_transform(datas['marital'])
datas['education'] = encoder.fit_transform(datas['education'])
datas['default'] = encoder.fit_transform(datas['default'])
datas['housing'] = encoder.fit_transform(datas['housing'])
datas['loan'] = encoder.fit_transform(datas['loan'])
datas['poutcome'] = encoder.fit_transform(datas['poutcome'])
datas['y'] = encoder.fit_transform(datas['y'])
datas['contact'] = encoder.fit_transform(datas['contact'])
datas['month'] = encoder.fit_transform(datas['month'])
```

datas.head()

	age	job	marital	education	default	balance	housing	loan	contact	day	month
0	58	4	1	2	0	2143	1	0	2	5	8
1	44	9	2	1	0	29	1	0	2	5	8
2	33	2	1	1	0	2	1	1	2	5	8
5	35	4	1	2	0	231	1	0	2	5	8
6	28	4	2	2	0	447	1	1	2	5	8
4											<b>&gt;</b>

del datas['default']

del datas['contact']

datas.head()

	age	job	marital	education	balance	housing	loan	day	month	duration	campai
0	58	4	1	2	2143	1	0	5	8	261	
1	44	9	2	1	29	1	0	5	8	151	
2	33	2	1	1	2	1	1	5	8	76	
5	35	4	1	2	231	1	0	5	8	139	
6	28	4	2	2	447	1	1	5	8	217	
4											<b>&gt;</b>

X= datas.iloc[:,0:14]

X[:]

```
age job marital education balance housing loan day month duration ca
y = datas.iloc[:,14]
y[:]
    0
            0
    1
             0
    2
            0
    45206
    45207
    45208
    45209
            0
    45210
    Name: y, Length: 43193, dtype: int64
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=22)
     TULIU U 1 L 1 LUII U U 11 U
X_train.shape, y_train.shape
    ((30235, 14), (30235,))
X_test.shape, y_test.shape
    ((12958, 14), (12958,))
svc_model = SVC()
svc_model.fit(X_train,y_train)
     ▼ SVC
     SVC()
predictions = svc_model.predict(X_test)
print(confusion_matrix(y_test,predictions))
     [[11498
     [ 1437
              16]]
accuracy_score(y_test, predictions)
    0.8885630498533724
accuracy_score(y_train, svc_model.predict(X_train))
    0.8820902927071275
print(classification_report(y_test,predictions))
                 precision recall f1-score support
              0
                      0.89
                               1.00
                                        0.94
                                              11505
                     0.70
                                        0.02
                                               1453
              1
                               0.01
                                                12958
                                        0.89
        accuracy
                     0.79
                            0.51
                                                12958
       macro avg
                                        0.48
    weighted avg
                     0.87
                               0.89
                                        0.84
                                              12958
```