Q. Write a program to implement Support Vector Machine Algorithm to solve classification problem on two different datasets individually.

For Dataset: Nissan-data.csv

Importing necessary libraries

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import classification_report, confusion_matrix,
accuracy_score
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler, LabelEncoder
```

Importing first dataset and head of the df

```
df = pd.read_csv('nissan-dataset.csv')
df.head()
```

```
   id            full_name   age      gender          model    color  \
0   1      Dominic Applin  42.0        Male          Quest     Mauv
1   2         Lenee Eteen  54.0  Polygender        R'nessa   Orange
2   3      Kendal Esselin  37.0        Male  March / Micra     Teal
3   4  Nehemiah Marvelley  55.0        Male         Gloria    Green
4   5    Domenic McGeouch  21.0        Male         Avenir    Khaki

   performance         km  condition     price
0        299.0  509305.0   very bad  40394.91
1        109.0  965853.0        old   8687.90
2         52.0  380906.0        bad  44705.31
3        336.0  573171.0  very good  32889.88
4          2.0  809470.0        old   6949.22

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 10 columns):

 #   Column        Non-Null Count   Dtype
---  -------       --------------   -------
 0   id            10000 non-null   int64
 1   full_name     9137 non-null    object
 2   age           9154 non-null    float64
 3   gender        9144 non-null    object
 4   model         9130 non-null    object
 5   color         9139 non-null    object
```

```
 6   performance  9143 non-null   float64
 7   km           9141 non-null   float64
 8   condition    9151 non-null   object
 9   price        9159 non-null   float64
dtypes: float64(4), int64(1), object(5)
memory usage: 781.4+ KB
```

```
df.isnull().sum()
```

```
id                0
full_name       843
age             846
gender          856
model           870
color           861
performance     857
km              859
condition       849
price           841
dtype: int64
```

```
df.dropna(inplace=True)
```

```
df.isnull().sum()
```

```
id              0
full_name       0
age             0
gender          0
model           0
color           0
performance     0
km              0
condition       0
price           0
dtype: int64
```

```
df.describe()
```

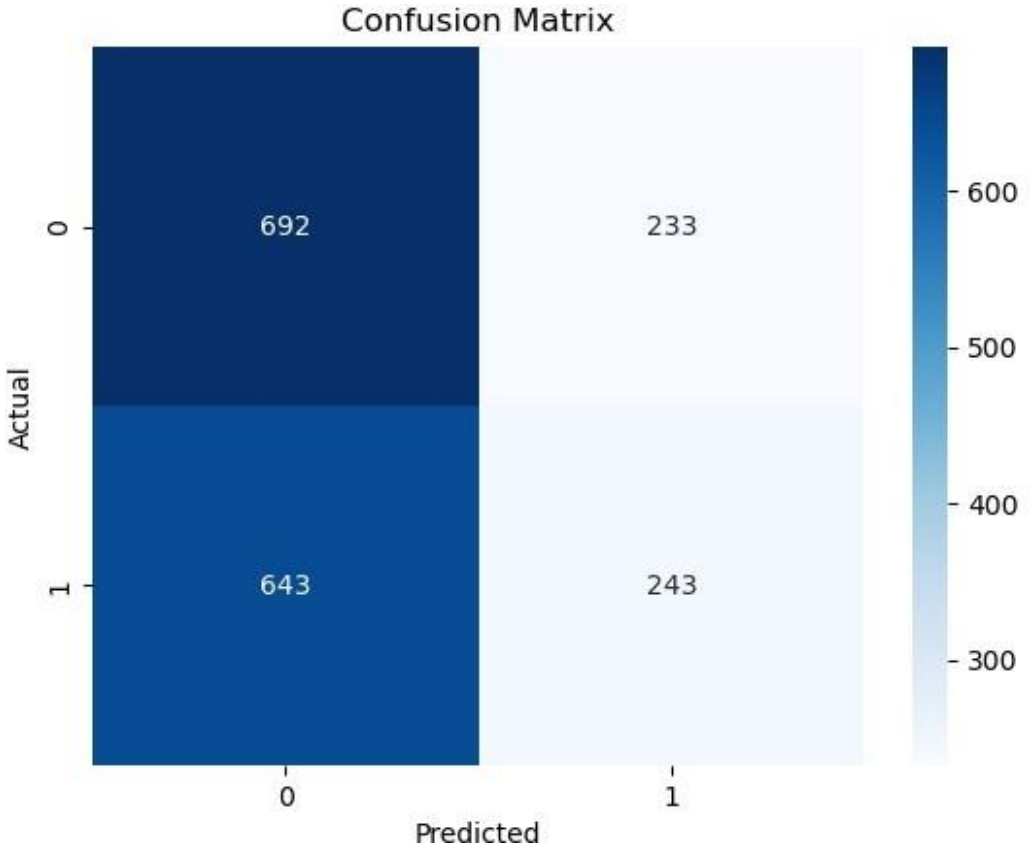
	id	age	performance	km	price
count	9053.000000	9053.000000	9053.000000	9053.000000	9053.000000
mean	4983.303435	43.080305	197.822379	503871.380979	27361.050557
std	2905.407440	14.783407	112.850882	284898.768044	13463.534018
min	1.000000	18.000000	0.000000	808.000000	5000.250000
25%	2470.000000	31.000000	102.000000	258709.000000	15960.500000
50%	4969.000000	43.000000	196.000000	502308.000000	27328.560000
75%	7510.000000	56.000000	296.000000	749260.000000	38385.630000
max	10000.000000	69.000000	399.000000	999915.000000	215674.780000

```
df['price'] = pd.to_numeric(df['price'], errors='coerce')

median_price = df['price'].median()
df['high_price'] = (df['price'] > median_price).astype(int)

df = df.drop(['id', 'full_name', 'price','gender'], axis=1)

df = pd.get_dummies(df, drop_first=True)

X = df.drop('high_price', axis=1)
y = df['high_price']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)

svm_model = SVC(kernel='linear')

svm_model.fit(X_train, y_train)

y_pred = svm_model.predict(X_test)

conf_matrix = confusion_matrix(y_test, y_pred)
class_report = classification_report(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)

print("Confusion Matrix:")
print(conf_matrix)

print("\nClassification Report:")
print(class_report)

print("\nAccuracy Score:")
print(accuracy)
```

```
Confusion Matrix:
[[692 233]
 [643 243]]

Classification Report:
              precision    recall  f1-score   support

           0       0.52      0.75      0.61       925
           1       0.51      0.27      0.36       886

    accuracy                           0.52      1811
   macro avg       0.51      0.51      0.48      1811
weighted avg       0.51      0.52      0.49      1811


Accuracy Score:
0.5162893429044727
```

```
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
```



For dataset: Plant growth data

```
df1 = pd.read_csv('plant_growth_data.csv')
df1.head()
```

	Soil_Type	Sunlight_Hours	Water_Frequency	Fertilizer_Type	Temperature	Humidity	Growth_Milestone
0	loam	5.192294	bi-weekly	chemical	31.719602	61.591861	0
1	sandy	4.033133	weekly	organic	28.919484	52.422276	1
2	loam	8.892769	bi-weekly	none	23.179059	44.660539	0
3	loam	8.241144	bi-weekly	none	18.465886	46.433227	0
4	sandy	8.374043	bi-weekly	organic	18.128741	63.625923	0

```
df1.isnull().sum()
```

```
Soil_Type            0
Sunlight_Hours       0
Water_Frequency      0
Fertilizer_Type      0
Temperature          0
Humidity             0
Growth_Milestone     0
dtype: int64
```

```
df1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 193 entries, 0 to 192
```

```
Data columns (total 7 columns):
 #   Column            Non-Null Count  Dtype  
---  ------            --------------  -----  
 0   Soil_Type         193 non-null    object 
 1   Sunlight_Hours    193 non-null    float64
 2   Water_Frequency   193 non-null    object 
 3   Fertilizer_Type   193 non-null    object 
 4   Temperature       193 non-null    float64
 5   Humidity          193 non-null    float64
 6   Growth_Milestone  193 non-null    int64  
dtypes: float64(3), int64(1), object(3)
memory usage: 10.7+ KB
```

```
df1.describe()
```

```
       Sunlight_Hours  Temperature    Humidity  Growth_Milestone
count      193.000000   193.000000  193.000000        193.000000
mean         6.826484    25.076087   58.098927          0.497409
std          1.599509     5.354170   12.631799          0.501294
min          4.033133    15.200000   30.567682          0.000000
25%          5.477000    20.637095   49.300000          0.000000
50%          6.833290    25.912336   59.182806          0.000000
75%          8.241144    29.757938   69.100000          1.000000
max          9.913903    34.810103   79.648240          1.000000
```

```
label_encoders = {}
categorical_columns = ['Soil_Type', 'Water_Frequency', 'Fertilizer_Type']
for column in categorical_columns:
    le = LabelEncoder()
    df1[column] = le.fit_transform(df1[column])
    label_encoders[column] = le
```

```
df1.head()
```

```
   Soil_Type  Sunlight_Hours  Water_Frequency  Fertilizer_Type   Temperature  \
0          1        5.192294                0                0    31.719602   
1          2        4.033133                2                2    28.919484   
2          1        8.892769                0                1    23.179059   
3          1        8.241144                0                1    18.465886   
4          2        8.374043                0                2    18.128741   

    Humidity  Growth_Milestone  
0  61.591861                 0  
1  52.422276                 1  
2  44.660539                 0  
3  46.433227                 0  
4  63.625923                 0  
```

```
X = df1.drop('Growth_Milestone', axis=1)
y = df1['Growth_Milestone']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)

scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

svm = SVC(kernel='linear')
svm.fit(X_train, y_train)
```

```
SVC(kernel='linear')

y_pred = svm.predict(X_test)

print("Classification Report:\n", classification_report(y_test, y_pred))
```

```
Classification Report:
               precision    recall  f1-score   support

           0       0.42      0.59      0.49        17
           1       0.53      0.36      0.43        22

    accuracy                           0.46        39
   macro avg       0.47      0.48      0.46        39
weighted avg       0.48      0.46      0.46        39
```

```
plt.figure(figsize=(8, 6))
sns.heatmap(confusion_matrix(y_test, y_pred), annot=True, fmt='d',
cmap='Blues')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.title('Confusion Matrix')
plt.show()
```