

Support Vector Machines

Aim:

Implement Support Vector Machines using dataset. Displaying it.

Algorithm:

- Step 1: Importing the dataset
- Step 2: Splitting the dataset into training and test samples.
- Step 3: Classifying the predictors and target
- Step 4: Initializing Support Vector Machines and fitting the training Data.
- Step 5: Predicting the classes for test set.
- Step 6: Attaching the predictions to test set for comparing.
- Step 7: Display the Accuracy of SVM of the Dataset and printing the Accuracy Metrics and Confusion Matrix.
- Step 8: Visualizing the Dataset with the color of red and Orange for a given Dataset.

Program:

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix, classification_report
from sklearn.preprocessing import LabelEncoder
import numpy as np
```

```

import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap

pd.options.mode.chained_assignment = None

data = pd.read_csv("apples-and-oranges.csv")
print(data.head(1))

training_set, test_set = train_test_split(data, test_size=0.2, random_state=1)
X_train = training_set.iloc[:, 0:2].values
Y_train = training_set.iloc[:, 2].values
X_test = test_set.iloc[:, 0:2].values
Y_test = test_set.iloc[:, 2].values

classifier = SVC(kernel='rbf', random_state=5)
classifier.fit(X_train, Y_train)

Y_pred = classifier.predict(X_test)
test_set["predictions"] = Y_pred

cm = ConfusionMatrix(Y_test, Y_pred)
accuracy = float(cm.diagonal().sum() / len(Y_test))
print("\nAccuracy of SVM for the Given Dataset :", accuracy)

le = LabelEncoder()
Y_train = le.fit_transform(Y_train)

classifier = SVC(kernel='rbf', random_state=1)
classifier.fit(X_train, Y_train)
print('Accuracy Metrics')
print('Classification_report (Y_test, Y_pred)')

```



```

print('Confusion Matrix')
print(confusion_Matrix(y_test, y_pred))

print('
plt.figure(figsize=(7,7))
X_set, y_set = X_train, y_train'

x1, x2 = np.meshgrid(np.arange(start=X_set[:,0].min()-1, stop=X_set[:,0].
max()+1, step=0.01), np.arange(start=X_set[:,1].min()-1, stop=X_set
plt.xlim(x1.min(), x1.max())
plt.ylim(x2.min(), x2.max())
for i, j in enumerate(np.unique(y_set)):
    plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1], c= ListedColormap
                (('red', 'orange'))(i), label = j)

plt.title('Apple vs Orange')
plt.xlabel('weight In Grams')
plt.ylabel('Size in cm')
plt.legend()
plt.show()

```

Result:

Thus the Implementation of Support Vector Machines using Dataset is Verified and Successfully executed.

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localhost:8888/notebooks/ML/exp9.ipynb

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Trusted Python 3

```
In [10]: import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix, classification_report
from sklearn.preprocessing import LabelEncoder
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap

In [11]: pd.options.mode.chained_assignment = None

In [12]: data = pd.read_csv(r"C:\Users\Smart\anaconda3\dataset\apples_and_oranges.csv")

In [13]: print(data.head())
training_set, test_set = train_test_split(data, test_size = 0.2, random_state = 1)

  Weight  Size  Class
0      69  4.39  orange
1      69  4.21  orange
2      65  4.09  orange
3      72  5.85   apple
4      67  4.70  orange

In [14]: X_train = training_set.iloc[:,0:2].values
```

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Run Code

```
In [14]: X_train = training_set.iloc[:,0:2].values
Y_train = training_set.iloc[:,2].values
X_test = test_set.iloc[:,0:2].values
Y_test = test_set.iloc[:,2].values

In [15]: classifier = SVC(kernel='rbf', random_state = 5)
classifier.fit(X_train,Y_train)

Out[15]: SVC(random_state=5)

In [16]: Y_pred = classifier.predict(X_test)
test_set["Predictions"] = Y_pred

In [17]: cm = confusion_matrix(Y_test,Y_pred)
accuracy = float(cm.diagonal().sum())/len(Y_test)
print("\nAccuracy Of SVM For The Given Dataset : ", accuracy)

Accuracy Of SVM For The Given Dataset : 0.375

In [18]: le = LabelEncoder()
Y_train = le.fit_transform(Y_train)

In [19]: classifier = SVC(kernel='rbf', random_state = 1)
```

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+ %> Run Code

```
accuracy = float(cm.diagonal().sum())/len(Y_test)
print("\nAccuracy Of SVM For The Given Dataset : ", accuracy)
```

Accuracy Of SVM For The Given Dataset : 0.375

```
In [18]: le = LabelEncoder()
Y_train = le.fit_transform(Y_train)
```

```
In [19]: classifier = SVC(kernel='rbf', random_state = 1)
classifier.fit(X_train,Y_train)
print('Accuracy Metrics')
print(classification_report(Y_test,Y_pred))
```

Accuracy Metrics

	precision	recall	f1-score	support
apple	0.38	1.00	0.55	3
orange	0.00	0.00	0.00	5
accuracy			0.38	8
macro avg	0.19	0.50	0.27	8
weighted avg	0.14	0.38	0.20	8

C:\Users\Smart\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1245: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use 'zero_division' parameter to control this beha

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In [20]:

```
print('Confusion Matrix')
print(confusion_matrix(Y_test, Y_pred))
```

Confusion Matrix

```
[[3 0]
 [5 0]]
```

In [21]:

```
plt.figure(figsize = (7,7))
X_set, y_set = X_train, Y_train
X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, 0].max() + 1, step = 0.01), np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, 1].max() + 1, step = 0.01))
plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(X1.shape), alpha = 0.75, cmap = ListedColormap(('red', 'orange')))
plt.xlim(X1.min(), X1.max())
plt.ylim(X2.min(), X2.max())
for i, j in enumerate(np.unique(y_set)):
    plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1], c = ListedColormap(('red', 'orange'))(i), label = j)
plt.title('Apples Vs Oranges')
plt.xlabel('Weight In Grams')
plt.ylabel('Size in cm')
plt.legend()
plt.show()
```

c argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.

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