

SVM Classification

18BCS063

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1. Use apple_and_oranges.csv dataset.

Import the dataset and divide the dataset into Training and Testing dataset.

- Apply SVM classifier (use given kernel type) to build model using Training Dataset.

Kernel Type = Radial basis function (RBF), gamma=0.8

Kernel Type = Linear

- Predict class label for data items in test Dataset.
- Print the Confusion Matrix and Calculate the accuracy of the predictions.
- Visualize the classifier results

```
In [80]: import pandas as pd
import numpy as np
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix
from sklearn.preprocessing import LabelEncoder
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
df = pd.read_csv("apples_and_oranges.csv")
```

```
In [81]: df.head()
```

```
Out[81]:
```

	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 [82]: df.info
```

```
Out[82]: <bound method DataFrame.info of
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
5      73  5.68   apple
6      70  5.56   apple
7      75  5.11   apple
8      74  5.36   apple
9      65  4.27  orange
```

```
In [83]: from sklearn.model_selection import train_test_split
```

```
In [84]: training_set, test_set = train_test_split(df, test_size = 0.2, random_state = 1)
```

```
In [61]: 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 [62]: classifier = SVC(kernel='rbf', random_state = 1, gamma=0.6)  
classifier.fit(X_train,Y_train)
```

```
Out[62]: SVC(gamma=0.6, random_state=1)
```

```
In [63]: Y_prediction = classifier.predict(X_test)
```

```
In [64]: test_set["Predictions"] = Y_prediction
```

```
<ipython-input-64-85ee06a09b3d>:1: SettingWithCopyWarning:  
A value is trying to be set on a copy of a slice from a DataFrame.  
Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy
```

```
test_set["Predictions"] = Y_prediction
```

```
In [65]: test_set
```

```
Out[65]:
```

```
In [65]: test_set
```

```
Out[65]:
```

	Weight	Size	Class	Predictions
2	65	4.09	orange	orange
31	66	4.68	orange	orange
3	72	5.85	apple	apple
21	70	4.83	orange	apple
27	70	4.22	orange	orange
29	71	5.26	apple	apple
22	69	4.61	orange	orange
39	73	5.03	apple	apple

```
In [66]: cm = confusion_matrix(Y_test,Y_prediction)  
accuracy = float(cm.diagonal().sum())/len(Y_test)  
print("Accuracy Of SVM For The Given Dataset : ", accuracy)
```

```
Accuracy Of SVM For The Given Dataset : 0.875
```

```
In [67]: le = LabelEncoder()  
Y_train = le.fit_transform(Y_train)
```

```
In [68]: classifier = SVC(kernel='rbf', random_state = 1, gamma=0.6)  
classifier.fit(X_train,Y_train)
```

```
Out[68]: SVC(gamma=0.6, random_state=1)
```

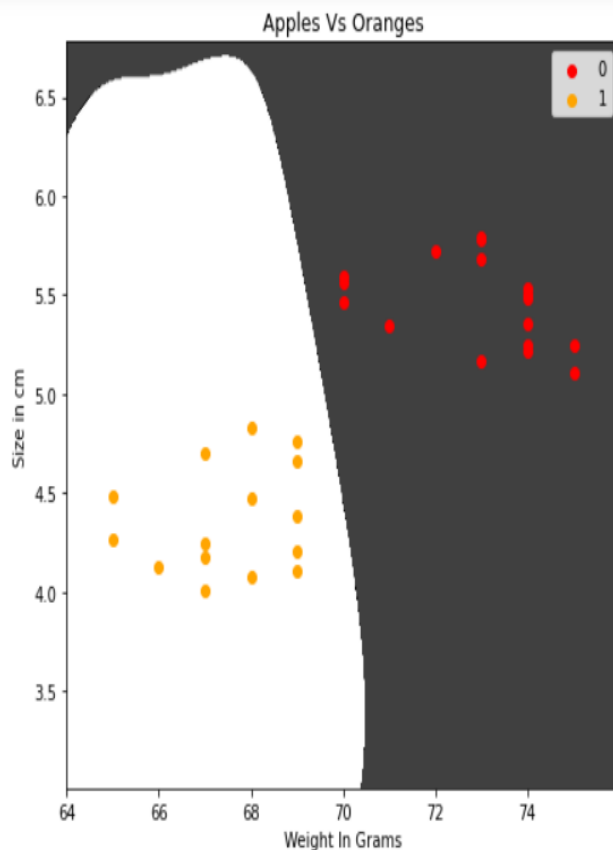
In [70]: #Actual Output

```
In [71]: 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(2))
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.

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.

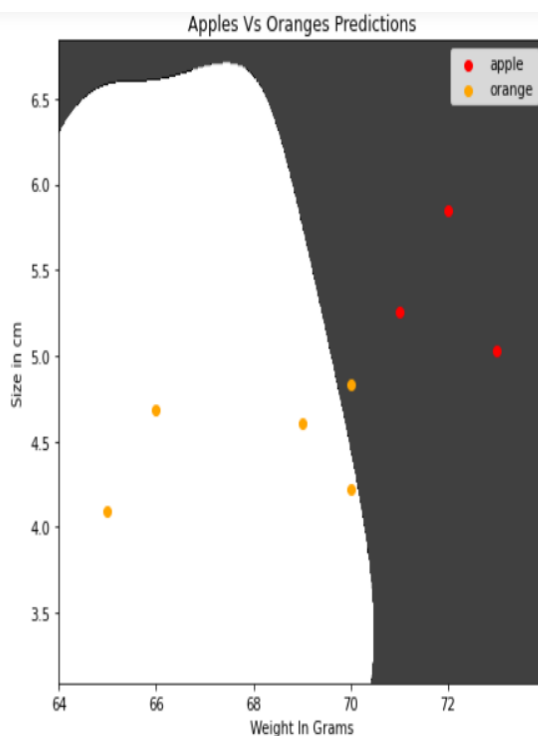
Apples Vs Oranges



```
In [72]: # Predicted Output
```

```
In [73]: plt.figure(figsize = (7,7))
X_set, y_set = X_test, Y_test
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 Predictions')
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.
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.



```
[85]: classifier2= SVC(kernel='linear', random_state = 1)
classifier2.fit(X_train,Y_train)
```

```
t[85]: SVC(kernel='linear', random_state=1)
```

```
[86]: Y_pred = classifier.predict(X_test)
```

```
In [86]: Y_pred = classifier.predict(X_test)
```

```
In [87]: test_set["Predictions"] = Y_pred
```

<ipython-input-87-946a65001e17>:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
test_set["Predictions"] = Y_pred
```

```
In [79]: test_set
```

Out[79]:

	Weight	Size	Class	Predictions
2	65	4.09	orange	1
31	66	4.68	orange	1
3	72	5.85	apple	0
21	70	4.83	orange	0
27	70	4.22	orange	1
29	71	5.26	apple	0
22	69	4.61	orange	1
39	73	5.03	apple	0

2. Use “iris.csv” dataset.

I. Import and Visualize the dataset.

```
In [7]: import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
from sklearn import datasets  
from sklearn import svm  
  
df=pd.read_csv("iris.csv")
```

```
In [8]: df.head()
```

Out[8]:

	sepal.length	sepal.width	petal.length	petal.width	variety
0	5.1	3.5	1.4	0.2	Setosa
1	4.9	3.0	1.4	0.2	Setosa
2	4.7	3.2	1.3	0.2	Setosa
3	4.6	3.1	1.5	0.2	Setosa
4	5.0	3.6	1.4	0.2	Setosa

```
In [9]: df
```

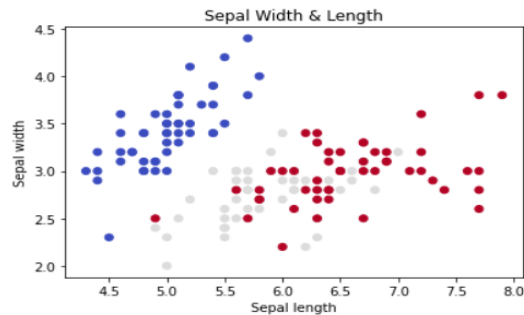
Out[9]:

	sepal.length	sepal.width	petal.length	petal.width	variety
0	5.1	3.5	1.4	0.2	Setosa
1	4.9	3.0	1.4	0.2	Setosa
2	4.7	3.2	1.3	0.2	Setosa
3	4.6	3.1	1.5	0.2	Setosa
4	5.0	3.6	1.4	0.2	Setosa
...

1. Visualizing the relationship between sepal width and target classes

```
In [10]: def visuvalize_sepal_data():
→iris = datasets.load_iris()
→X = iris.data[:, :2]
→y = iris.target
→plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.coolwarm)
→plt.xlabel('Sepal length')
→plt.ylabel('Sepal width')
→plt.title('Sepal width & Length')
→plt.show()

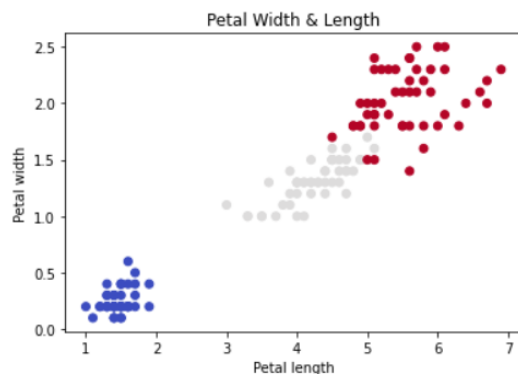
visuvalize_sepal_data()
```



2. Visualizing the relationship between petal width and target classes

```
In [11]: def visuvalize_petal_data():
→iris = datasets.load_iris()
→X = iris.data[:, 2:]
→y = iris.target
→plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.coolwarm)
→plt.xlabel('Petal length')
→plt.ylabel('Petal width')
→plt.title('Petal width & Length')
→plt.show()

visuvalize_petal_data()
```



II. Apply SVM (use given kernel type) for prediction.

1. Model given Kernel SVM classifier using Iris Sepal features and Visualizing the modeled SVM classifiers with Iris Sepal features

Kernel Type = Radial basis function (RBF), gamma=0.8

Kernel Type = Linear

Kernel Type = Polynomial, Degree = 3

```
In [12]: iris = datasets.load_iris()
X = iris.data[:, :2]
y = iris.target
C = 1.0
svc = svm.SVC(kernel='linear', C=C).fit(X, y)
lin_svc = svm.LinearSVC(C=C).fit(X, y)
rbf_svc = svm.SVC(kernel='rbf', gamma=0.7, C=C).fit(X, y)
poly_svc = svm.SVC(kernel='poly', degree=3, C=C).fit(X, y)
```

C:\Users\A B Charan\anaconda3\lib\site-packages\sklearn\svm_base.py:976: ConvergenceWarning: Liblinear failed to converge, increase the number of iterations.
warnings.warn("Liblinear failed to converge, increase "

```
In [13]: h = .02
x_min, x_max = X[:, 0].min() - 1, X[:, 0].max() + 1
y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
                    np.arange(y_min, y_max, h))

titles = ['SVC with linear kernel',
          'LinearSVC (linear kernel)',
          'SVC with RBF kernel',
          'SVC with polynomial (degree 3) kernel']

for i, clf in enumerate((svc, lin_svc, rbf_svc, poly_svc)):
    plt.subplot(2, 2, i + 1)
    plt.subplots_adjust(wspace=0.4, hspace=0.4)

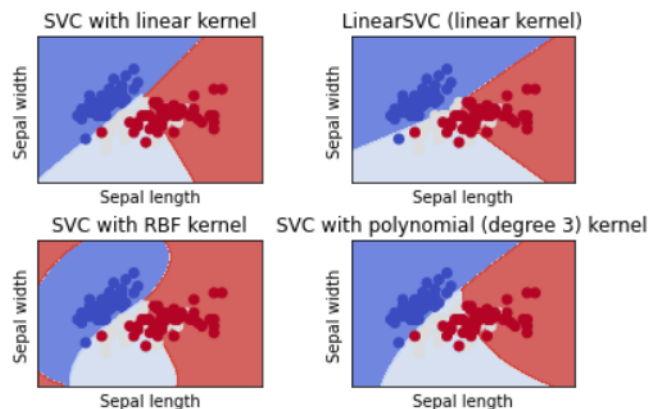
    Z = clf.predict(np.c_[xx.ravel(), yy.ravel()])

    Z = Z.reshape(xx.shape)
    plt.contourf(xx, yy, Z, cmap=plt.cm.coolwarm, alpha=0.8)
```

```
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.coolwarm, alpha=0.8)

plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.coolwarm)
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.ylim(yy.min(), yy.max())
plt.xticks(())
plt.yticks(())
plt.title(titles[i])

plt.show()
```



2. Model given Kernel SVM classifier using Iris Sepal features and Visualizing the modeled SVM classifiers with Iris petal features

```
In [14]: iris = datasets.load_iris()
X = iris.data[:, 2:] # we only take the last two features.
y = iris.target
C = 1.0 # SVM regularization parameter

# SVC with linear kernel
svc = svm.SVC(kernel='linear', C=C).fit(X, y)
# LinearSVC (linear kernel)
lin_svc = svm.LinearSVC(C=C).fit(X, y)
# SVC with RBF kernel
rbf_svc = svm.SVC(kernel='rbf', gamma=0.7, C=C).fit(X, y)
# SVC with polynomial (degree 3) kernel
poly_svc = svm.SVC(kernel='poly', degree=3, C=C).fit(X, y)

In [15]: h = .02 # step size in the mesh
# create a mesh to plot in
x_min, x_max = X[:, 0].min() - 1, X[:, 0].max() + 1
y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
                    np.arange(y_min, y_max, h))
# title for the plots
titles = ['SVC with linear kernel',
        'LinearSVC (linear kernel)',
        'SVC with RBF kernel',
        'SVC with polynomial (degree 3) kernel']

for i, clf in enumerate((svc, lin_svc, rbf_svc, poly_svc)):
    # Plot the decision boundary. For that, we will assign a color to each
    # point in the mesh [x_min, x_max]x[y_min, y_max].
    plt.subplot(2, 2, i + 1)
    plt.subplots_adjust(wspace=0.4, hspace=0.4)

    Z = clf.predict(np.c_[xx.ravel(), yy.ravel()])
```

```
Z = clf.predict(np.c_[xx.ravel(), yy.ravel()])

# Put the result into a color plot
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.coolwarm, alpha=0.8)

# Plot also the training points
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.coolwarm)
plt.xlabel('Petal length')
plt.ylabel('Petal width')
plt.xlim(xx.min(), xx.max())
plt.ylim(yy.min(), yy.max())
plt.xticks(())
plt.yticks(())
plt.title(titles[i])

plt.show()
```

