import numpy as np

```
#To implement SVM using any inbuild and external data set.
```

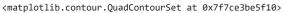
```
import matplotlib.pyplot as plt
from sklearn import svm, datasets

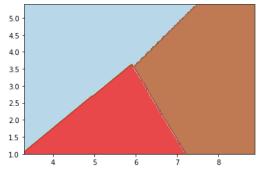
# import some data to play with
iris = datasets.load_iris()
X = iris.data[:, :2] # we only take the first two features. We could
# avoid this ugly slicing by using a two-dim dataset
y = iris.target
```

```
# we create an instance of SVM and fit out data. We do not scale our
# data since we want to plot the support vectors
# SVM regularization parameter
C = 1.0
svc = svm.SVC(kernel='linear',C=1,gamma='auto').fit(X, y)
```

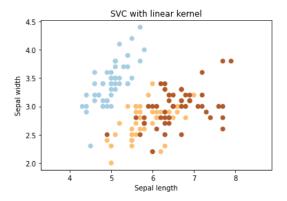
```
# create a mesh to plot in
x_min, x_max = X[:, 0].min() - 1, X[:, 0].max() + 1
y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
h = (x_max / x_min)/100
xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
np.arange(y_min, y_max, h))
```

```
plt.subplot(1, 1, 1)
Z = svc.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.Paired, alpha=0.8)
```



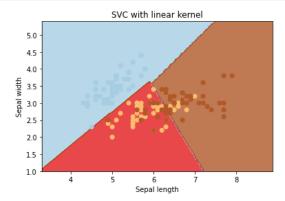


```
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.title('SVC with linear kernel')
plt.show()
```



```
plt.subplot(1, 1, 1)
Z = svc.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.Paired, alpha=0.8)
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)
plt.xlabel('Sepal length')
```

```
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.title('SVC with linear kernel')
plt.show()
```

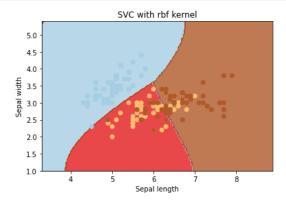


#Use SVM rbf kernel

```
svc = svm.SVC(kernel='rbf', C=1,gamma='auto').fit(X, y)
```

```
# 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
h = (x_max / x_min)/100
xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
np.arange(y_min, y_max, h))
```

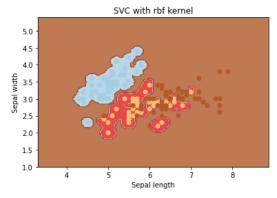
```
plt.subplot(1, 1, 1)
Z = svc.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.Paired, alpha=0.8)
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.title('SVC with rbf kernel')
plt.show()
```



```
svc = svm.SVC(kernel='rbf', C=1,gamma=10).fit(X, y)
```

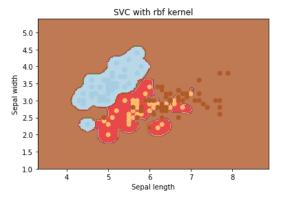
```
plt.subplot(1, 1, 1)
Z = svc.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.Paired, alpha=0.8)
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.title('SVC with rbf kernel')
plt.show()
```

```
SVC with rbf kernel
        5.0
        4.5
        4.0
      width
        3.5
      Sepal v
        2.5
svc = svm.SVC(kernel='rbf', C=1,gamma=100).fit(X, y)
        1.0 -
plt.subplot(1, 1, 1)
Z = svc.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.Paired, alpha=0.8)
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.title('SVC with rbf kernel')
plt.show()
```



```
svc = svm.SVC(kernel='rbf', C=1,gamma=50).fit(X, y)
```

```
plt.subplot(1, 1, 1)
Z = svc.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.contourf(xx, yy, Z, cmap=plt.cm.Paired, alpha=0.8)
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)
plt.xlabel('Sepal length')
plt.ylabel('Sepal width')
plt.xlim(xx.min(), xx.max())
plt.title('SVC with rbf kernel')
plt.show()
```



```
import pandas as pd
data = pd.read_csv("apples_and_oranges.csv")

from sklearn.model_selection import train_test_split
training_set, test_set = train_test_split(data, test_size = 0.2, random_state = 1)
```

plt.legend() plt.show()

```
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
from sklearn.svm import SVC
classifier = SVC(kernel='rbf', random_state = 1)
classifier.fit(X_train,Y_train)
     SVC(random_state=1)
Y_pred = classifier.predict(X_test)
test_set["Predictions"] = Y_pred
from sklearn.metrics import confusion_matrix
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
{\it from \ sklearn.preprocessing \ import \ Label Encoder}
le = LabelEncoder()
Y_train = le.fit_transform(Y_train)
from sklearn.svm import SVC
classifier = SVC(kernel='rbf', random_state = 1)
classifier.fit(X_train,Y_train)
     SVC(random_state=1)
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
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[:, 0].min()} - 1
plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(X1.shape), alpha = 0.75, cmap = ListedColormap(
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')
```

```
*c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in
import numpy as np
 import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
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, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = X_set[:, \ 0].max() + 1, \ stop = 0.01), np.arange(start = X_set[:, \ 1].min() - 1, \ stop = 
plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(X1.shape),alpha = 0.75, cmap = ListedColormap((
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 *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in

