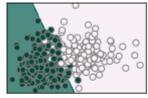
```
#Importing the necessary packages and libaries
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn import svm, datasets
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
breast_cancer= datasets.load_breast_cancer()
#Store variables as target y and the first two features as X (sepal length and sepal width of the iris flow
X = breast cancer.data[:,:2]
y = breast_cancer.target
X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.8, random_state = 0)
linear = svm.SVC(kernel='linear', C=1, decision_function_shape='ovo').fit(X_train, y_train)
rbf = svm.SVC(kernel='rbf', gamma=1, C=1, decision_function_shape='ovo').fit(X_train, y_train)
poly = svm.SVC(kernel='poly', degree=3, C=1, decision_function_shape='ovo').fit(X_train, y_train)
sig = svm.SVC(kernel='sigmoid', C=1, decision_function_shape='ovo').fit(X_train, y_train)
#stepsize in the mesh, it alters the accuracy of the plotprint
#to better understand it, just play with the value, change it and print it
h = .01
#create the mesh
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))
# create the title that will be shown on the plot
titles = ['Linear kernel', 'RBF kernel', 'Polynomial kernel', 'Sigmoid kernel']
for i, clf in enumerate((linear, rbf, poly, sig)):
#defines how many plots: 2 rows, 2columns=> leading to 4 plots
plt.subplot(2, 2, i + 1) #i+1 is the index
#space between plots
plt.subplots_adjust(wspace=0.4, hspace=0.4)
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.PuBuGn, alpha=0.7)
# Plot also the training points
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.PuBuGn, edgecolors='grey')
plt.xlim(xx.min(), xx.max())
plt.ylim(yy.min(), yy.max())
plt.xticks(())
plt.yticks(())
plt.title(titles[i])
plt.show()
linear_pred = linear.predict(X_test)
poly pred = poly.predict(X test)
rbf_pred = rbf.predict(X_test)
sig_pred = sig.predict(X_test)
```

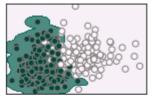
Гэ

Linear kernel





RBF kernel



Sigmoid kernel

# retrieve the accuracy and print it for all 4 kernel functions
accuracy\_lin = linear.score(X\_test, y\_test)
accuracy\_poly = poly.score(X\_test, y\_test)
accuracy\_rbf = rbf.score(X\_test, y\_test)
accuracy\_sig = sig.score(X\_test, y\_test)
print("Accuracy Linear Kernel:", accuracy\_lin)

Accuracy Linear Kernel: 0.8947368421052632

print("Accuracy Polynomial Kernel:", accuracy\_poly)
print("Accuracy Radial Basis Kernel:", accuracy\_rbf)
print("Accuracy Sigmoid Kernel:", accuracy\_sig)
# creating a confusion matrix
cm\_lin = confusion\_matrix(y\_test, linear\_pred)
cm\_poly = confusion\_matrix(y\_test, poly\_pred)
cm\_rbf = confusion\_matrix(y\_test, rbf\_pred)
cm\_sig = confusion\_matrix(y\_test, sig\_pred)
print(cm\_lin)
print(cm\_poly)
print(cm\_rbf)
print(cm\_sig)

Accuracy Polynomial Kernel: 0.8947368421052632 Accuracy Radial Basis Kernel: 0.8421052631578947 Accuracy Sigmoid Kernel: 0.5877192982456141

[[40 7]

[562]]

[[41 6]

[ 6 61]]

[[37 10]

[8 59]]

[[ 0 47]

[ 0 67]]