

```
In [67]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LogisticRegression
from sklearn.datasets import make_blobs
from sklearn import metrics
from sklearn.model_selection import train_test_split
import seaborn as sns; sns.set()
from scipy import stats
from matplotlib.colors import ListedColormap
from sklearn.preprocessing import MinMaxScaler, StandardScaler
from sklearn.datasets import load_breast_cancer
from sklearn.decomposition import PCA
from sklearn.svm import SVC, SVR
```

Problem 1

```
In [68]: canc_feats = pd.DataFrame(load_breast_cancer().data)
canc_labels = np.reshape(load_breast_cancer().target, (np.size(load_breast_cancer().data)))
canc_scaled = StandardScaler().fit_transform(canc_feats)
canc_data = np.concatenate([canc_scaled, canc_labels], axis=1)
```

```
In [69]: linear_model = SVC(kernel = "linear", C = 1000)
poly_model = SVC(kernel = "poly", C = 1000)
rbf_model = SVC(kernel = "rbf", C = 1000)
sigmoid_model = SVC(kernel = "sigmoid", C = 1000)

canc_train, canc_test = train_test_split(canc_data, train_size = 0.8, test_size = 0.2)

canc_train_results = canc_train[:, -1]
canc_train = np.delete(canc_train, canc_train.shape[1] - 1, -1)
canc_test_results = canc_test[:, -1]
canc_test = np.delete(canc_test, canc_test.shape[1] - 1, -1)
```

```
In [70]: pca_data = []
pca_val = []
for n in range(0, canc_train.shape[1]+1):
    pca = PCA(n_components=n)
    pca_data.append(pca.fit_transform(canc_train))
    pca_val.append(pca.fit_transform(canc_test))
```

```

In [71]: max_test_acc_linear = 0
n_max_test_linear = 0
linear_acc = []

max_test_acc_poly = 0
n_max_test_poly = 0
poly_acc = []

max_test_acc_rbf = 0
n_max_test_rbf = 0
rbf_acc = []

max_test_acc_sigmoid = 0
n_max_test_sigmoid = 0
sigmoid_acc = []

for n in range(1, canc_train.shape[1]+1):

    linear_model.fit(pca_data[n], canc_train_results)
    linear_acc.append(linear_model.score(pca_val[n], canc_test_results))

    poly_model.fit(pca_data[n], canc_train_results)
    poly_acc.append(poly_model.score(pca_val[n], canc_test_results))

    rbf_model.fit(pca_data[n], canc_train_results)
    rbf_acc.append(rbf_model.score(pca_val[n], canc_test_results))

    sigmoid_model.fit(pca_data[n], canc_train_results)
    sigmoid_acc.append(sigmoid_model.score(pca_val[n], canc_test_results))

    if linear_model.score(pca_val[n], canc_test_results) > max_test_acc_linear:
        max_test_acc_linear = linear_model.score(pca_val[n], canc_test_results)
        n_max_test_linear = n

    if poly_model.score(pca_val[n], canc_test_results) > max_test_acc_poly:
        max_test_acc_poly = poly_model.score(pca_val[n], canc_test_results)
        n_max_test_poly = n

    if rbf_model.score(pca_val[n], canc_test_results) > max_test_acc_rbf:
        max_test_acc_rbf = rbf_model.score(pca_val[n], canc_test_results)
        n_max_test_rbf = n

    if sigmoid_model.score(pca_val[n], canc_test_results) > max_test_acc_sigmoid:
        max_test_acc_sigmoid = sigmoid_model.score(pca_val[n], canc_test_results)
        n_max_test_sigmoid = n

    print(f"""
Linear Model:
N = {n}
Training Accuracy: {linear_model.score(pca_data[n], canc_train_results)}
Test Accuracy: {linear_model.score(pca_val[n], canc_test_results)}
""")
    print(f"""
Poly Model:
N = {n}

```

```

Training Accuracy: {poly_model.score(pca_data[n], canc_train_results)}
Test Accuracy: {poly_model.score(pca_val[n], canc_test_results)}
"""
    print(f"""
RBF Model:
N = {n}
Training Accuracy: {rbf_model.score(pca_data[n], canc_train_results)}
Test Accuracy: {rbf_model.score(pca_val[n], canc_test_results)}
"""
        )
    print(f"""
Sigmoid Model:
N = {n}
Training Accuracy: {sigmoid_model.score(pca_data[n], canc_train_results)}
Test Accuracy: {sigmoid_model.score(pca_val[n], canc_test_results)}
"""
        )

    print(f"""
Max Test Accuracy for Linear Model: {max_test_acc_linear}
N: {n_max_test_linear}

Max Test Accuracy for Linear Model: {max_test_acc_poly}
N: {n_max_test_poly}

Max Test Accuracy for Linear Model: {max_test_acc_rbf}
N: {n_max_test_rbf}

Max Test Accuracy for Linear Model: {max_test_acc_sigmoid}
N: {n_max_test_sigmoid}
""")

```

Test Accuracy: 0.8333333333333334

Linear Model:

N = 2

Training Accuracy: 0.9582417582417583

Test Accuracy: 0.9122807017543859

Poly Model:

N = 2

Training Accuracy: 0.9362637362637363

Test Accuracy: 0.9035087719298246

RBF Model:

N = 2

Training Accuracy: 0.9582417582417583

Test Accuracy: 0.9122807017543859

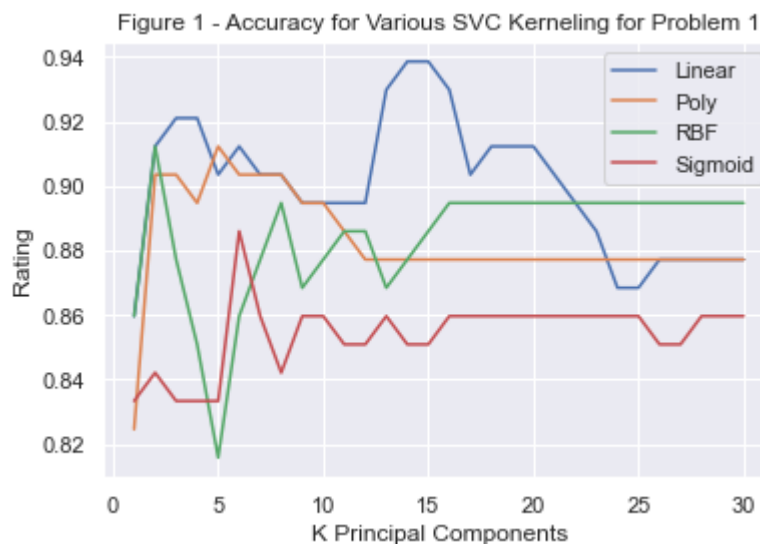
```

In [72]: linear_model.fit(pca_data[n_max_test_linear], canc_train_results)
poly_model.fit(pca_data[n_max_test_poly], canc_train_results)
rbf_model.fit(pca_data[n_max_test_rbf], canc_train_results)
sigmoid_model.fit(pca_data[n_max_test_sigmoid], canc_train_results)

xx = range(1, canc_train.shape[1]+1)
plot1 = plt.figure(1)
plt.plot(xx, linear_acc)
plt.plot(xx, poly_acc)
plt.plot(xx, rbf_acc)
plt.plot(xx, sigmoid_acc)
plt.legend(['Linear', 'Poly', 'RBF', 'Sigmoid'], loc='upper right')
plt.xlabel('K Principal Components')
plt.ylabel('Rating')
plt.title("Figure 1 - Accuracy for Various SVC Kerneling for Problem 1")

```

Out[72]: Text(0.5, 1.0, 'Figure 1 - Accuracy for Various SVC Kerneling for Problem 1')



Problem 2

```
In [73]: ▶ feat_list = ['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom',  
house = pd.DataFrame(pd.read_csv(r'C:\Users\ccm51\Documents\ECGR_4105\Housing  
house.replace("yes", 1, inplace=True)  
house.replace("no", 0, inplace=True)  
house_scaled = StandardScaler().fit_transform(house)  
house_scaled.shape
```

Out[73]: (545, 12)

```
In [74]: ▶ house_results = house_scaled[:, -1]  
house_results  
house_scaled = np.delete(house_scaled, house_scaled.shape[1]-1, -1)  
print(house_results.shape, house_scaled.shape)
```

(545,) (545, 11)

```
In [75]: ▶ house_train, house_test = train_test_split(house_scaled, train_size = 0.8, te  
  
house_train_results = house_train[:, -1]  
house_train = np.delete(house_train, house_train.shape[1] - 1, -1)  
house_test_results = house_test[:, -1]  
house_test = np.delete(house_test, house_test.shape[1] - 1, -1)
```

```
In [76]: ▶ model_rbf = SVR(kernel = "rbf", gamma = 1, C = 100)  
model_lin = SVR(kernel = "linear", C = 100)  
model_poly = SVR(kernel = "poly", C = 100, degree = 2)  
results_rbf = model_rbf.fit(house_train, house_train_results).predict(house_t  
results_lin = model_lin.fit(house_train, house_train_results).predict(house_t  
results_poly = model_poly.fit(house_train, house_train_results).predict(house
```

```
In [80]: ▶ pca_data = []  
pca_val = []  
for n in range(1, house_train.shape[1]+1):  
    pca = PCA(n_components=n)  
    pca_data.append(pca.fit_transform(house_train))  
    pca_val.append(pca.fit_transform(house_test))
```

```

In [78]: max_test_acc_linear = 0
n_max_test_linear = 0
linear_acc = []

max_test_acc_poly = 0
n_max_test_poly = 0
poly_acc = []

max_test_acc_rbf = 0
n_max_test_rbf = 0
rbf_acc = []

for n in range(0, house_train.shape[1]):

    model_lin.fit(pca_data[n], house_train_results)
    linear_acc.append(model_lin.score(pca_val[n], house_test_results))

    model_poly.fit(pca_data[n], house_train_results)
    poly_acc.append(model_poly.score(pca_val[n], house_test_results))

    model_rbf.fit(pca_data[n], house_train_results)
    rbf_acc.append(model_rbf.score(pca_val[n], house_test_results))

    if model_lin.score(pca_val[n], house_test_results) > max_test_acc_linear:
        max_test_acc_linear = model_lin.score(pca_val[n], house_test_results)
        n_max_test_linear = n

    if model_poly.score(pca_val[n], house_test_results) > max_test_acc_poly:
        max_test_acc_poly = model_poly.score(pca_val[n], house_test_results)
        n_max_test_poly = n

    if model_rbf.score(pca_val[n], house_test_results) > max_test_acc_rbf:
        max_test_acc_rbf = model_rbf.score(pca_val[n], house_test_results)
        n_max_test_rbf = n

    print(f"""
Linear Model:
N = {n+1}
Training Accuracy: {model_lin.score(pca_data[n], house_train_results)}
Test Accuracy: {model_lin.score(pca_val[n], house_test_results)}
""")

    print(f"""
Poly Model:
N = {n+1}
Training Accuracy: {model_poly.score(pca_data[n], house_train_results)}
Test Accuracy: {model_poly.score(pca_val[n], house_test_results)}
""")

    print(f"""
RBF Model:
N = {n+1}
Training Accuracy: {model_rbf.score(pca_data[n], house_train_results)}
Test Accuracy: {model_rbf.score(pca_val[n], house_test_results)}
""")

    print(f"""
Training Accuracy: {model_lin.score(pca_data[n], house_train_results)}
Test Accuracy: -0.24414973846607269
""")

```

RBF Model:

N = 6

Training Accuracy: 0.7901233275645467

Test Accuracy: -0.08904963198070681

Linear Model:

N = 7

Training Accuracy: -0.19768799121326763

Test Accuracy: -0.24086842321479862

Poly Model:

N = 7

Training Accuracy: -0.19034776101664796

Test Accuracy: -0.24553891145104356

```
In [84]: xx = range(1, house_train.shape[1]+1)
plot1 = plt.figure(2)
plt.plot(xx, linear_acc)
plt.plot(xx, poly_acc)
plt.plot(xx, rbf_acc)
plt.legend(['Linear', 'Poly', 'RBF'], loc='lower right')
plt.xlabel('K Principal Components')
plt.ylabel('Rating')
plt.title("Figure 2 - Accuracy for Various SVR Kerneling for Problem 2")
```

Out[84]: Text(0.5, 1.0, 'Figure 2 - Accuracy for Various SVR Kerneling for Problem 2')

