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In [1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns

from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from sklearn.decomposition import IncrementalPCA
from sklearn.metrics import classification_report
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
from sklearn.metrics import precision_recall_curve
from sklearn import metrics

In [2]: dataset = load_breast_cancer()

X = dataset.data

Y = dataset.target

In [3]: X_train, X_test, y_train, y_test = train_test_split(X, Y, train_size = 0.8, test_size = 0.2, random_state = 0)

sc = StandardScaler()

Xsc_train = sc.fit_transform(X_train)
Xsc_test = sc.transform(X_test)

In [4]: #sns.scatterplot(x=Xsc_train[:,0],y=Xsc_train[:,1],hue=y_train)

In [5]: model = PCA()

model.fit(Xsc_train)

#explained_variance = np.cumsum(model.explained_variance_ratio_)
#plt.plot(explained_variance)

Out[5]: PCA()

In [32]: PCA_train = IncrementalPCA(n_components = 15)
df_train = PCA_train.fit_transform(Xsc_train)
df_test = PCA_train.transform(Xsc_test)

In [33]: svc_lin = SVC(kernel='linear', C=100)
svc_rbf = SVC(kernel='rbf', C=100)
svc_ply = SVC(kernel='poly', C=100)

svc_lin = svc_lin.fit(df_train, y_train)
svc_rbf = svc_rbf.fit(df_train, y_train)
svc_ply = svc_ply.fit(df_train, y_train)

In [34]: predicted_lin = svc_lin.predict(df_test)
predicted_rbf = svc_rbf.predict(df_test)
predicted_ply = svc_ply.predict(df_test)

print('Linear:',classification_report(y_test,predicted_lin, digits=4))
print('RBF:',classification_report(y_test,predicted_rbf, digits=4))
print('Poly:',classification_report(y_test,predicted_ply, digits=4))
```

Linear:		precision	recall	f1-score	support
	0	0.9375	0.9574	0.9474	47
	1	0.9697	0.9552	0.9624	67
	accuracy			0.9561	114
	macro avg	0.9536	0.9563	0.9549	114
	weighted avg	0.9564	0.9561	0.9562	114
RBF:		precision	recall	f1-score	support
	0	0.9038	1.0000	0.9495	47
	1	1.0000	0.9254	0.9612	67
	accuracy			0.9561	114
	macro avg	0.9519	0.9627	0.9554	114
	weighted avg	0.9604	0.9561	0.9564	114
Poly:		precision	recall	f1-score	support
	0	0.9783	0.9574	0.9677	47
	1	0.9706	0.9851	0.9778	67
	accuracy			0.9737	114
	macro avg	0.9744	0.9713	0.9728	114
	weighted avg	0.9738	0.9737	0.9736	114

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In [35]: print('Accuracy:', accuracy_score(y_test, predicted_lin))
print('Precision:', metrics.precision_score(y_test, predicted_lin))
print('Recall:', metrics.recall_score(y_test, predicted_lin))
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Accuracy: 0.956140350877193
Precision: 0.9696969696969697
Recall: 0.9552238805970149
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In [36]: print('Accuracy:', accuracy_score(y_test, predicted_rbf))
print('Precision:', metrics.precision_score(y_test, predicted_rbf))
print('Recall:', metrics.recall_score(y_test, predicted_rbf))
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Accuracy: 0.956140350877193
Precision: 1.0
Recall: 0.9253731343283582
```

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In [37]: print('Accuracy:', accuracy_score(y_test, predicted_ply))
print('Precision:', metrics.precision_score(y_test, predicted_ply))
print('Recall:', metrics.recall_score(y_test, predicted_ply))
```

```
Accuracy: 0.9736842105263158
Precision: 0.9705882352941176
Recall: 0.9850746268656716
```

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In [ ]: #Q2
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In [57]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
import warnings

from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from sklearn.decomposition import IncrementalPCA
from sklearn.metrics import classification_report
from sklearn.svm import SVR
from sklearn.metrics import accuracy_score
from sklearn.metrics import precision_recall_curve
from sklearn import metrics
from sklearn.metrics import mean_squared_error
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In [58]: warnings.filterwarnings('ignore')
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In [59]: dataset = pd.read_csv('Housing.csv')
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df = dataset.drop('furnishingstatus', axis=1)
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In [60]: svar_list = ['mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning', 'prefarea']
def binary_mapping(x):
    return x.map({'yes':1, 'no':0})
df[svar_list] = df[svar_list].apply(binary_mapping)

df.head()
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Out[60]:
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	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	pr
0	13300000	7420	4	2	3	1	0	0	0	1	2	
1	12250000	8960	4	4	4	1	0	0	0	1	3	
2	12250000	9960	3	2	2	1	0	1	0	0	2	
3	12215000	7500	4	2	2	1	0	1	0	1	3	
4	11410000	7420	4	1	2	1	1	1	0	1	2	

```
In [61]: X = df[['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning']]
y = dataset['price']

X.shape
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Out[61]: (545, 11)
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In [62]: y.shape
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Out[62]: (545,)
```

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In [67]: X_train, X_test, y_train, y_test = train_test_split(X, y, train_size = 0.8, test_size = 0.2, random_state = 0)

sc = StandardScaler()

Xsc_train = sc.fit_transform(X_train)
Xsc_test = sc.transform(X_test)
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In [81]: svr_rbf = SVR(kernel='rbf', C=1e6, gamma=0.1)
svr_lin = SVR(kernel='linear', C=1e6)
svr_ply = SVR(kernel='poly', C=1e6, degree=2)

y_rbf = svr_rbf.fit(Xsc_train, y_train).predict(Xsc_test)
y_lin = svr_lin.fit(Xsc_train, y_train).predict(Xsc_test)
y_ply = svr_ply.fit(Xsc_train, y_train).predict(Xsc_test)

print('Loss:', mean_squared_error(y_test, y_lin), mean_squared_error(y_test, y_rbf), mean_squared_error(y_test, y_ply))

Loss: 947860338641.9976 1279027590158.7593 1622321986946.383
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In [82]: model = PCA()

model.fit(Xsc_train)
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Out[82]: PCA()
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In [83]: PCA_train = IncrementalPCA(n_components = 11)
df_train = PCA_train.fit_transform(Xsc_train)
df_test = PCA_train.transform(Xsc_test)
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In [86]: y_rbf_pca = svr_rbf.fit(df_train, y_train).predict(X_test)
y_lin_pca = svr_lin.fit(df_train, y_train).predict(X_test)
y_ply_pca = svr_ply.fit(df_train, y_train).predict(X_test)
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In [90]: print('Loss:', mean_squared_error(y_test, y_lin_pca), mean_squared_error(y_test, y_rbf_pca), mean_squared_error(y_test, y_ply_pca))

Loss: 2.662630291102947e+19 3086028271056.9614 4.5372558100900655e+25
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In [ ]:
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