21BAI1533 PCA and SVM

PCA

```
In [83]: import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
In [84]: url = "https://archive.ics.uci.edu/ml/machine-learning-databases/ir
          names = ['sepal-length', 'sepal-width', 'petal-length', 'petal-widt
          df=pd.read_csv(url, names=names)
In [85]: df.head()
Out[85]:
             sepal-length sepal-width petal-length petal-width
                                                          Class
           0
                    5.1
                               3.5
                                         1.4
                                                   0.2 Iris-setosa
                    4.9
                               3.0
                                         1.4
                                                   0.2 Iris-setosa
           2
                    4.7
                               3.2
                                         1.3
                                                   0.2 Iris-setosa
           3
                    4.6
                               3.1
                                         1.5
                                                   0.2 Iris-setosa
                    5.0
                               3.6
                                         1.4
                                                   0.2 Iris-setosa
In [86]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 150 entries, 0 to 149
          Data columns (total 5 columns):
           #
               Column
                               Non-Null Count
                                                Dtype
               sepal-length 150 non-null
                                                 float64
               sepal-width
                               150 non-null
                                                 float64
           1
           2
               petal-length 150 non-null
                                                 float64
           3
               petal-width
                               150 non-null
                                                 float64
                               150 non-null
                                                object
               Class
```

dtypes: float64(4), object(1)

memory usage: 6.0+ KB

```
In [87]: df.describe()
```

sonal langth sonal width notal langth notal width

Out [87]:

	sepai-length	sepai-width	petai-iength	petal-width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
In [88]: x=df.drop('Class', 1)
y=df['Class']
```

/var/folders/72/3kxng2yd5yn203b626zw3kpc0000gn/T/ipykernel_17046/3 513278735.py:1: FutureWarning: In a future version of pandas all a rguments of DataFrame.drop except for the argument 'labels' will be keyword-only.

x=df.drop('Class', 1)

```
In [89]: from sklearn.model_selection import train_test_split
```

```
In [90]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

```
In [91]: from sklearn.preprocessing import StandardScaler
```

```
In [92]: sc=StandardScaler()
    x_train=sc.fit_transform(x_train)
    x_test=sc.transform(x_test)
```

```
In [93]: from sklearn.decomposition import PCA
```

```
In [94]: pca = PCA()
```

```
In [95]: x_train=pca.fit_transform(x_train)
x_test=pca.transform(x_test)
```

```
In [96]: var=pca.explained_variance_ratio_
var
```

Out[96]: array([0.73053554, 0.22968726, 0.03446579, 0.00531141])

```
In [97]: from sklearn.ensemble import RandomForestClassifier
 In [98]: classifier = RandomForestClassifier(max depth=2, random state=0)
          classifier.fit(x_train, y_train)
 Out[98]: RandomForestClassifier(max depth=2, random state=0)
In [100]: y_pred = classifier.predict(x_test)
In [101]: | from sklearn.metrics import confusion_matrix
          from sklearn.metrics import accuracy_score
In [102]: pca = PCA(n_components=1)
          x_train = pca.fit_transform(x_train)
          x_test = pca.transform(x_test)
In [103]: | cm = confusion_matrix(y_test, y_pred)
          print(cm)
          print(accuracy_score(y_test, y_pred))
          [[15 0
                   01
           [ 0 10 8]
           [ 0 0 12]]
          0.82222222222222
In [107]: pca = PCA()
          X_train = pca.fit_transform(x_train)
          X_test = pca.transform(x_test)
In [108]:
          cm = confusion_matrix(y_test, y_pred)
          print(accuracy_score(y_test, y_pred))
          [[15 0
                   01
           [ 0 10 8]
           [ 0 0 12]]
          0.82222222222222
          SVM
In [156]: | from sklearn.datasets import load_breast_cancer
In [157]: | cancer = load_breast_cancer()
```

In [158]: col_names = list(cancer.feature_names)
 col_names.append('target')
 df = pd.DataFrame(np.c_[cancer.data, cancer.target], columns=col_na
 df.head()

Out[158]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	symr
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0

5 rows × 31 columns

In [159]: print(cancer.target_names)

['malignant' 'benign']

In [160]: df.describe()

Out[160]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	r conc
count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.00
mean	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	80.0
std	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.07
min	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.00
25%	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.02
50%	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.06
75 %	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.13
max	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.42

8 rows × 31 columns

In [161]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 31 columns):

#	Column	Non-Null Count	Dtype
0	mean radius	569 non-null	float64
1	mean texture	569 non-null	float64
2	mean perimeter	569 non-null	float64
3	mean area	569 non-null	float64
4	mean smoothness	569 non-null	float64
5	mean compactness	569 non-null	float64
6	mean concavity	569 non-null	float64
7	mean concave points	569 non-null	float64
8	mean symmetry	569 non-null	float64
9	mean fractal dimension	569 non-null	float64
10	radius error	569 non-null	float64
11	texture error	569 non-null	float64
12	perimeter error	569 non-null	float64
13	area error	569 non-null	float64
14	smoothness error	569 non-null	float64
15	compactness error	569 non-null	float64
16	concavity error	569 non-null	float64
17	concave points error	569 non-null	float64
18	symmetry error	569 non-null	float64
19	fractal dimension error	569 non-null	float64
20	worst radius	569 non-null	float64
21	worst texture	569 non-null	float64
22	worst perimeter	569 non-null	float64
23	worst area	569 non-null	float64
24	worst smoothness	569 non-null	float64
25	worst compactness	569 non-null	float64
26	worst concavity	569 non-null	float64
27	worst concave points	569 non-null	float64
28	worst symmetry	569 non-null	float64
29	worst fractal dimension	569 non-null	float64
30	target	569 non-null	float64
	(1 (64/24)		

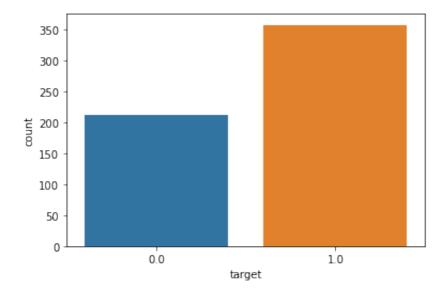
dtypes: float64(31)
memory usage: 137.9 KB

```
In [162]: df.columns
```

```
In [163]: #sns.pairplot(df,hue='target')
```

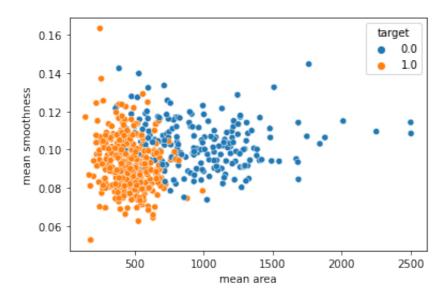
```
In [164]: sns.countplot(x=df['target'], label = "Count")
```

Out[164]: <AxesSubplot:xlabel='target', ylabel='count'>



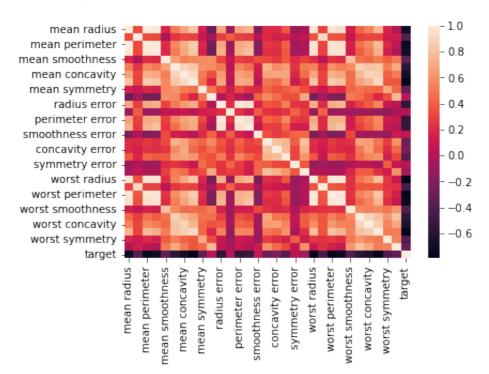
In [165]: sns.scatterplot(x = 'mean area', y = 'mean smoothness', hue = 'targ

Out[165]: <AxesSubplot:xlabel='mean area', ylabel='mean smoothness'>



In [166]: sns.heatmap(df.corr())

Out[166]: <AxesSubplot:>



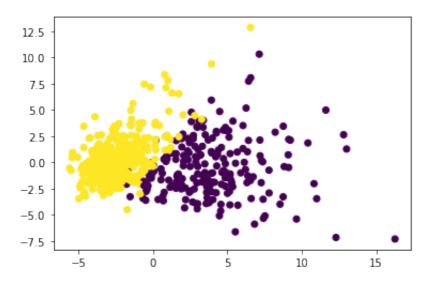
In [167]: from sklearn.model_selection import cross_val_score, train_test_spl
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler, MinMaxScaler

In [168]: x=df.drop('target', axis=1)
y=df.target

```
In [169]: x.shape
Out[169]: (569, 30)
In [170]: | y.shape
Out[170]: (569,)
In [171]: |,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=42)
In [179]: | sc=StandardScaler()
In [180]: sc.fit(df)
          /Users/zeelmehta/opt/anaconda3/lib/python3.9/site-packages/sklearn
          /utils/validation.py:1688: FutureWarning: Feature names only suppo
          rt names that are all strings. Got feature names with dtypes: ['st
          r', 'str ']. An error will be raised in 1.2.
            warnings.warn(
Out[180]: StandardScaler()
In [182]: | scaled=sc.transform(df)
          /Users/zeelmehta/opt/anaconda3/lib/python3.9/site-packages/sklearn
          /utils/validation.py:1688: FutureWarning: Feature names only suppo
          rt names that are all strings. Got feature names with dtypes: ['st
          r', 'str ']. An error will be raised in 1.2.
            warnings.warn(
In [184]: from sklearn.decomposition import PCA
In [186]: pca=PCA(n_components=2)
In [188]: pca.fit(scaled)
Out[188]: PCA(n_components=2)
In [189]: xpca=pca.transform(scaled)
In [190]: |scaled.shape
Out[190]: (569, 31)
In [191]: xpca.shape
Out[191]: (569, 2)
```

```
In [194]: plt.scatter(xpca[:,0],xpca[:,1],c=cancer['target'])
```

Out[194]: <matplotlib.collections.PathCollection at 0x2e8bc9fa0>



```
In [55]: pca.components_
```

Out[55]: array([[-0.21890244, -0.10372458, -0.22753729, -0.22099499, -0.142 58969, -0.23928535, -0.25840048, -0.26085376, -0.13816696, -0.06436335. -0.20597878, -0.01742803, -0.21132592, -0.20286964, -0.01453145, -0.17039345, -0.15358979, -0.1834174, -0.04249842, -0.10256832, -0.22799663, -0.10446933, -0.23663968, -0.22487053, -0.12795256, -0.21009588, -0.22876753, -0.25088597, -0.12290456, -0.131783941. [0.23385713, 0.05970609, 0.21518136, 0.23107671, -0.186 11302, -0.15189161, -0.06016536, 0.0347675, -0.19034877, -0.36657547, 0.10555215, -0.08997968, 0.08945723, 0.15229263, -0.20443045, -0.2327159 , -0.19720728 , -0.13032156 , -0.183848 , -0.28009203, 0.21986638, 0.0454673, 0.19987843, 0.21935186, -0.172 30435, -0.14359317, -0.09796411, 0.00825724, -0.14188335, -0.275 3394711)

In [195]: from sklearn.svm import SVC

```
In [196]:
svm=SVC()
```

In [197]: svm.fit(x_train,y_train)

/Users/zeelmehta/opt/anaconda3/lib/python3.9/site-packages/sklearn/utils/validation.py:1688: FutureWarning: Feature names only support names that are all strings. Got feature names with dtypes: ['str_']. An error will be raised in 1.2. warnings.warn(

Out[197]: SVC()

In [198]: pred=svm.predict(x_test)

/Users/zeelmehta/opt/anaconda3/lib/python3.9/site-packages/sklearn/utils/validation.py:1688: FutureWarning: Feature names only support names that are all strings. Got feature names with dtypes: ['str_']. An error will be raised in 1.2.

warnings.warn(

In [199]: from sklearn.metrics import classification_report,confusion_matrix

In [200]: print(confusion_matrix(y_test,pred))

[[52 11] [0 108]]

In [201]: print(classification_report(y_test,pred))

	precision	recision recall		support	
0.0 1.0	1.00 0.91	0.83 1.00	0.90 0.95	63 108	
accuracy macro avg weighted avg	0.95 0.94	0.91 0.94	0.94 0.93 0.93	171 171 171	