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# PRINCIPAL COMPONENT ANALYSIS
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# importing the scikit-learn.dataset library

import pandas as pd

from sklearn import datasets

import matplotlib.pyplot as plt

from sklearn.metrics import accuracy_score, confusion_matrix, classification_report

from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.decomposition import PCA

from sklearn.model_selection import GridSearchCV

from sklearn.preprocessing import StandardScaler

from sklearn.model_selection import train_test_split

# %%

# generating the dataset for 1000 samples with linear relation from make_regression() fucntion of skearn

x, y = datasets.make_classification(n_samples=1000, # number of samples
                                   n_features=4) # number of features

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# visualizing each feature
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```

fig, axs = plt.subplots(2, 2)
axs[0, 0].scatter(x[:, 0], y)
axs[0, 0].set_title('feature 1')
axs[0, 1].scatter(x[:, 1], y)
axs[0, 1].set_title('feature 2')
axs[1, 0].scatter(x[:, 2], y)
axs[1, 0].set_title('feature 3')
axs[1, 1].scatter(x[:, 3], y)
axs[1, 1].set_title('feature 4')

# Hide x labels and tick labels for top plots and y ticks for right plots.
for ax in axs.flat:
    ax.label_outer()

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# creating training and test datasets with train_test_split func. of sklearn.model_selection

train_x, test_x, train_y, test_y = train_test_split(x, y, test_size=0.2)

print("train data size:", len(train_x)) # 800

print("test data size:", len(test_x)) # 200

# %%

# Logistic Regression

# fitting the the logistic regression model

LR = LogisticRegression()

LR.fit(train_x, train_y)

# model prediction

LR_predict = LR.predict(test_x)

# model accuracy

print("Linear Regression Model accuracy: ",
      accuracy_score(test_y, LR_predict)*100)

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# Decision Tree
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```
# fitting the the decision tree model
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```
decision_tree = DecisionTreeClassifier()
```

```
decision_tree.fit(train_x, train_y)
```

```
# model prediction
```

```
decision_predict = decision_tree.predict(test_x)
```

```
# model accuracy
```

```
print("Decision Tree Model accuracy: ",  
      accuracy_score(test_y, decision_predict)*100)
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# Support Vector Machines (SVM)
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# fitting the SVM model
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```
svm = SVC()
```

```
svm.fit(train_x, train_y)
```

```
# model prediction
```

```
svm_predict = svm.predict(test_x)
```

```
# model accuracy
```

```
print("SVM Model accuracy: ",  
      accuracy_score(test_y, svm_predict)*100)
```

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# %%  
  
# Random Forest Classifier  
  
# fitting the Random Forest model  
  
rf = RandomForestClassifier(n_estimators=800, max_depth=5)  
  
rf.fit(train_x, train_y)  
  
# model prediction  
  
rf_predict = rf.predict(test_x)  
  
# model accuracy  
  
print("Random Forest Model accuracy: ",  
      accuracy_score(test_y, rf_predict)*100)  
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# K-Nearest Neighbours (KNN)  
  
# fitting the KNN model  
  
knn = KNeighborsClassifier(n_neighbors=8)  
  
knn.fit(train_x, train_y)  
  
# model prediction  
  
knn_predict = knn.predict(test_x)  
  
# model accuracy  
  
print("KNN Model accuracy: ",  
      accuracy_score(test_y, knn_predict)*100)  
  
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```
# scaling our data
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```
scaler = StandardScaler()
X_train, X_test, y_train, y_test = train_test_split(
    x, y, test_size=0.3, random_state=42)
```

```
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

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# Principal Component Analysis (PCA)
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```
pca = PCA(n_components=3)
scaler = StandardScaler()
```

```
X_train = pca.fit_transform(X_train)
X_test = pca.transform(X_test)
```

```
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

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```
# plotting the data
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```
plt.figure(figsize=(8, 6))
plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap='plasma')
plt.xlabel('First principal component')
plt.ylabel('Second Principal Component')
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```
# creating a function to find the model accuracy
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```
def print_score(clf, X_train, y_train, X_test, y_test, train=True):
    if train:
        pred = clf.predict(X_train)
        clf_report = pd.DataFrame(classification_report(
            y_train, pred, output_dict=True))
        print("Train Result:\n=====")
        print(f"Accuracy Score: {accuracy_score(y_train, pred) * 100:.2f}%")
```

```

print("_____")
print(f"CLASSIFICATION REPORT:\n{clf_report}")
print("_____")
print(f"Confusion Matrix: \n {confusion_matrix(y_train, pred)}\n")

elif train == False:
    pred = clf.predict(X_test)
    clf_report = pd.DataFrame(
        classification_report(y_test, pred, output_dict=True))
    print("Test Result:\n=====")
    print(f"Accuracy Score: {accuracy_score(y_test, pred) * 100:.2f}%")
    print("_____")
    print(f"CLASSIFICATION REPORT:\n{clf_report}")
    print("_____")
    print(f"Confusion Matrix: \n {confusion_matrix(y_test, pred)}\n")
    cm = confusion_matrix(y_test, pred)
    fig, ax = plt.subplots(figsize=(8, 8))
    ax.imshow(cm)
    ax.grid(False)
    ax.xaxis.set(ticks=(0, 1), ticklabels=('Predicted 0s', 'Predicted 1s'))
    ax.yaxis.set(ticks=(0, 1), ticklabels=('Actual 0s', 'Actual 1s'))
    ax.set_ylim(1.5, -0.5)
    for i in range(2):
        for j in range(2):
            ax.text(j, i, cm[i, j], ha='center', va='center', color='red')
    plt.show()

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param_grid = {'C': [0.01, 0.1, 0.5, 1, 10, 100],
              'gamma': [1, 0.75, 0.5, 0.25, 0.1, 0.01, 0.001],
              'kernel': ['rbf', 'poly', 'linear']}

```

```

grid = GridSearchCV(SVC(), param_grid, verbose=1, cv=5)
grid.fit(X_train, y_train)
best_params = grid.best_params_
print(f"Best params: {best_params}")

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svm_clf = SVC(**best_params)
svm_clf.fit(X_train, y_train)

```

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print_score(svm_clf, X_train, y_train, X_test, y_test, train=True)
print_score(svm_clf, X_train, y_train, X_test, y_test, train=False)

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# THE END  
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