```
# Importing the Libraries
import numpy as np
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn import svm
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import accuracy_score
from sklearn.metrics import accuracy_score
import pandas as pd
# Data Collection
data_set = pd.read_csv('/content/diabetes.csv')
data_set
\rightarrow
          Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
                                                                                                                    扁
      0
                    6
                                                                      33.6
                           148
                                           72
                                                         35
                                                                   0
                                                                                              0.627
                                                                                                      50
                                                                                                               1
                                                                                                                    1
       1
                    1
                            85
                                           66
                                                         29
                                                                      26.6
                                                                                              0.351
                                                                   0
                                                                                                      31
                                                                                                               0
       2
                    8
                           183
                                                                      23.3
                                           64
                                                          0
                                                                                              0.672
                                                                                                      32
                                                                                                               1
       3
                    1
                                                                                                      21
                                                                                                               0
                            89
                                           66
                                                         23
                                                                  94
                                                                      28.1
                                                                                              0.167
       4
                    0
                           137
                                           40
                                                         35
                                                                 168 43.1
                                                                                              2.288
                                                                                                      33
                                                                                                               1
      763
                   10
                           101
                                           76
                                                         48
                                                                 180
                                                                      32.9
                                                                                              0.171
                                                                                                      63
                                                                                                               0
      764
                    2
                           122
                                           70
                                                         27
                                                                      36.8
                                                                                              0.340
                                                                                                      27
                                                                                                               0
      765
                    5
                           121
                                           72
                                                         23
                                                                 112 26.2
                                                                                              0.245
                                                                                                      30
                                                                                                               0
      766
                    1
                           126
                                           60
                                                          0
                                                                   0
                                                                      30.1
                                                                                              0.349
                                                                                                      47
                                                                                                               1
      767
                    1
                            93
                                           70
                                                         31
                                                                      30.4
                                                                                              0.315
                                                                                                      23
                                                                                                               0
     768 rows × 9 columns
             Generate code with data set
                                        View recommended plots
                                                                    New interactive sheet
 Next steps:
# Data Preparation
y = data_set['Outcome']
x = data_set.drop('Outcome', axis =1)
# Standardization and Scaling
scaler = StandardScaler()
scaler.fit(x)
standardized_data = scaler.transform(x)
print (standardized_data)
    1.4259954 ]
      [-0.84488505 -1.12339636 -0.16054575 ... -0.68442195 -0.36506078
       -0.19067191]
      [ \ 1.23388019 \ \ 1.94372388 \ \ -0.26394125 \ \dots \ \ -1.10325546 \ \ 0.60439732
       -0.10558415]
                   [ 0.3429808
       -0.27575966]
      [-0.84488505 0.1597866 -0.47073225 ... -0.24020459 -0.37110101
        1.17073215]
```

```
[-0.84488505 \ -0.8730192 \quad 0.04624525 \ \dots \ -0.20212881 \ -0.47378505
       -0.87137393]]
x = standardized_data
y = data_set['Outcome']
# Data Splitting
x_train, x_test, y_train, y_test, = train_test_split(x, y, test_size = 0.2, stratify= y, random_state=2)
# 1. Support Vector Machine (SVM)
# Import the necessary module
from sklearn import svm
# Initialize an empty dictionary called 'results' to store the accuracy scores.
results = {}
classifier = svm.SVC(kernel='linear')
classifier.fit(x train, y train)
y_pred = classifier.predict(x_test)
results['SVM'] = accuracy_score(y_test, y_pred)
# 2. Random Forest Classifier
classifier = RandomForestClassifier(n estimators=100, random state=2)
classifier.fit(x_train, y_train)
y_pred = classifier.predict(x_test)
results['Random Forest'] = accuracy_score(y_test, y_pred)
# 3. Logistic Regression
classifier = LogisticRegression(random_state=2)
classifier.fit(x_train, y_train)
y_pred = classifier.predict(x_test)
results['Logistic Regression'] = accuracy_score(y_test, y_pred)
# 4. Decision Tree Classifier
classifier = DecisionTreeClassifier(random_state=2)
classifier.fit(x train, y train)
y_pred = classifier.predict(x_test)
results['Decision Tree'] = accuracy_score(y_test, y_pred)
# 5. K-Nearest Neighbors (KNN)
classifier = KNeighborsClassifier(n_neighbors=5) # Experiment with different 'n_neighbors'
classifier.fit(x_train, y_train)
y_pred = classifier.predict(x_test)
results['KNN'] = accuracy_score(y_test, y_pred)
# 6. Neural Network (MLPClassifier)
classifier = MLPClassifier(hidden_layer_sizes=(100,), max_iter=500, random_state=42)
classifier.fit(x_train, y_train)
y_pred = classifier.predict(x_test)
results['Neural Network'] = accuracy_score(y_test, y_pred)
# Print Results
print("Accuracy Scores:")
for model, accuracy in results.items():
    print(f"{model}: {accuracy:.4f}")
# Find the best model
best_model = max(results, key=results.get)
print(f"\nBest Model: {best_model} with accuracy {results[best_model]:.4f}")
→ Accuracy Scores:
     SVM: 0.7727
     Random Forest: 0.7273
     Logistic Regression: 0.7597
     Decision Tree: 0.6948
     KNN: 0.7208
     Neural Network: 0.7403
```

```
Best Model: SVM with accuracy 0.7727
#Final Model Testing
input_data =(35,186,84,42,89,35,46,0.286)
input_data_as_numpy_array = np.asarray(input_data)
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
std_data = scaler.transform(input_data_reshaped)
prediction = classifier.predict(std_data)
print(prediction)
if(prediction[0]==0):
  print('The person is not diabetic\n')
else:
  print('The person is diabetic\n')
₹
    [0]
     The person is not diabetic
     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid feature n
       warnings.warn(
```