untitled4

June 28, 2024

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[6]: import numpy as np
     import pandas as pd
     from sklearn.model_selection import train_test_split
     from sklearn.neighbors import KNeighborsClassifier
     from sklearn.linear_model import LogisticRegression, LinearRegression
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.ensemble import RandomForestClassifier
     from sklearn.metrics import accuracy_score, mean_squared_error
     from google.colab import drive
     # Mount Google Drive
     drive.mount('/content/drive')
     # Load the cancer dataset from Google Drive
     data_cancer = pd.read_csv("/content/drive/MyDrive/cancer_dataset/Cancer_Data.
      ⇔csv")
     print(data_cancer.head())
     # Hypothetical continuous target variable 'Tumor_Size'
     # Assuming 'Tumor_Size' is in the dataset
     {\it \# Here, I will create a synthetic 'Tumor\_Size' column for demonstration purposes}
     np.random.seed(42)
     data_cancer['Tumor_Size'] = np.random.uniform(1.0, 5.0, size=len(data_cancer))
     # Select features and target for regression
     X = data_cancer[['radius_mean', 'area_mean']]
     y_classification = data_cancer['diagnosis']
     y_regression = data_cancer['Tumor_Size']
     # Split data into training and testing sets
     X train, X test, y train classification, y test classification = 1
      -train_test_split(X, y_classification, test_size=0.2, random_state=42)
     _, _, y_train_regression, y_test_regression = train_test_split(X, y_regression, u
      →test_size=0.2, random_state=42)
     # K-NN Classifier
     k = 3
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knn = KNeighborsClassifier(n_neighbors=k)
knn.fit(X_train, y_train_classification)
# Logistic Regression Classifier
logistic_model = LogisticRegression()
logistic_model.fit(X_train, y_train_classification)
# Decision Tree Classifier
decision_tree = DecisionTreeClassifier(random_state=42)
decision_tree.fit(X_train, y_train_classification)
# Random Forest Classifier
random_forest = RandomForestClassifier(n_estimators=100, random_state=42)
random_forest.fit(X_train, y_train_classification)
# Linear Regression Model
linear_model = LinearRegression()
linear_model.fit(X_train, y_train_regression)
# Get user input
radius_mean = float(input("Enter radius mean: "))
area_mean = float(input("Enter area mean: "))
# Prepare new data for prediction
new_data = np.array([[radius_mean, area_mean]])
# K-NN Prediction
knn_prediction = knn.predict(new_data)
# Logistic Regression Prediction
logistic_prediction = logistic_model.predict(new_data)
# Decision Tree Prediction
tree_prediction = decision_tree.predict(new_data)
# Random Forest Prediction
forest_prediction = random_forest.predict(new_data)
# Linear Regression Prediction
linear_prediction = linear_model.predict(new_data)
# Output predictions
if knn_prediction[0] == 'M':
   print("K-NN prediction: cancer is present")
else:
   print("K-NN prediction: there is no cancer")
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if logistic_prediction[0] == 'M':
   print("Logistic Regression prediction: cancer is present")
else:
   print("Logistic Regression prediction: there is no cancer")
if tree_prediction[0] == 'M':
   print("Decision Tree prediction: cancer is present")
else:
   print("Decision Tree prediction: there is no cancer")
if forest prediction[0] == 'M':
   print("Random Forest prediction: cancer is present")
else:
   print("Random Forest prediction: there is no cancer")
print(f"Linear Regression prediction for Tumor_Size: {linear_prediction[0]:.
 92f}")
# Evaluate models
y_pred_knn = knn.predict(X_test)
knn accuracy = accuracy score(y test classification, y pred knn)
print(f"K-NN Model accuracy: {knn_accuracy:.2f}")
y_pred_logistic = logistic_model.predict(X_test)
logistic_accuracy = accuracy_score(y_test_classification, y_pred_logistic)
print(f"Logistic Regression Model accuracy: {logistic_accuracy:.2f}")
y_pred_tree = decision_tree.predict(X_test)
tree_accuracy = accuracy_score(y_test_classification, y_pred_tree)
print(f"Decision Tree Model accuracy: {tree_accuracy:.2f}")
y_pred_forest = random_forest.predict(X_test)
forest_accuracy = accuracy_score(y_test_classification, y_pred_forest)
print(f"Random Forest Model accuracy: {forest_accuracy:.2f}")
y_pred_linear = linear_model.predict(X_test)
linear_mse = mean_squared_error(y_test_regression, y_pred_linear)
print(f"Linear Regression Model MSE: {linear_mse:.2f}")
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Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	\
0	842302	M	17.99	10.38	122.80	1001.0	
1	842517	M	20.57	17.77	132.90	1326.0	
2	84300903	M	19.69	21.25	130.00	1203.0	
3	84348301	M	11.42	20.38	77.58	386.1	
4	84358402	M	20.29	14.34	135.10	1297.0	

```
smoothness_mean
                    compactness_mean concavity_mean concave points_mean \
0
           0.11840
                              0.27760
                                                0.3001
                                                                     0.14710
1
           0.08474
                              0.07864
                                                0.0869
                                                                     0.07017
2
           0.10960
                              0.15990
                                                0.1974
                                                                     0.12790
3
           0.14250
                              0.28390
                                                0.2414
                                                                     0.10520
4
           0.10030
                              0.13280
                                                0.1980
                                                                     0.10430
                                                    smoothness worst \
      texture_worst perimeter_worst area_worst
0
              17.33
                               184.60
                                            2019.0
                                                               0.1622
              23.41
                                                               0.1238
1
                               158.80
                                            1956.0
2
              25.53
                                            1709.0
                                                               0.1444
                               152.50
3
              26.50
                                98.87
                                            567.7
                                                               0.2098
                               152.20
                                            1575.0
                                                               0.1374
4
              16.67
                                        concave points_worst symmetry_worst
   compactness_worst
                       concavity_worst
0
              0.6656
                                0.7119
                                                       0.2654
                                                                        0.4601
              0.1866
                                0.2416
                                                       0.1860
                                                                        0.2750
1
2
              0.4245
                                0.4504
                                                       0.2430
                                                                        0.3613
3
              0.8663
                                0.6869
                                                       0.2575
                                                                        0.6638
              0.2050
4
                                0.4000
                                                       0.1625
                                                                        0.2364
   fractal_dimension_worst Unnamed: 32
0
                    0.11890
                                     NaN
1
                    0.08902
                                     NaN
2
                    0.08758
                                     NaN
3
                    0.17300
                                     NaN
4
                    0.07678
                                     NaN
```

[5 rows x 33 columns]
Enter radius mean: 17.99
Enter area mean: 1001.0

K-NN prediction: cancer is present

Logistic Regression prediction: cancer is present

Decision Tree prediction: cancer is present Random Forest prediction: cancer is present

Linear Regression prediction for Tumor Size: 2.94

K-NN Model accuracy: 0.89

Logistic Regression Model accuracy: 0.92

Decision Tree Model accuracy: 0.83 Random Forest Model accuracy: 0.91 Linear Regression Model MSE: 1.57

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but KNeighborsClassifier was fitted with feature names

warnings.warn(

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does

not have valid feature names, but LogisticRegression was fitted with feature names

warnings.warn(

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but DecisionTreeClassifier was fitted with feature names

warnings.warn(

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but RandomForestClassifier was fitted with feature names

warnings.warn(

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(