

Assignment_3

October 9, 2024

1 IMPLEMENTATION OF K-NEAREST NEIGHBORS ALGORITHM.

1.1 Importing Libraries and Dataset

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

```
[2]: df = pd.read_csv('diabetes.csv')
```

```
[4]: df
```

```
[4]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	
..	
763	10	101	76	48	180	32.9	
764	2	122	70	27	0	36.8	
765	5	121	72	23	112	26.2	
766	1	126	60	0	0	30.1	
767	1	93	70	31	0	30.4	

	Pedigree	Age	Outcome
0	0.627	50	1
1	0.351	31	0
2	0.672	32	1
3	0.167	21	0
4	2.288	33	1
..
763	0.171	63	0
764	0.340	27	0
765	0.245	30	0
766	0.349	47	1

```
767      0.315    23      0
```

```
[768 rows x 9 columns]
```

```
[5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Pregnancies     768 non-null   int64
1   Glucose         768 non-null   int64
2   BloodPressure   768 non-null   int64
3   SkinThickness   768 non-null   int64
4   Insulin         768 non-null   int64
5   BMI             768 non-null   float64
6   Pedigree        768 non-null   float64
7   Age            768 non-null   int64
8   Outcome         768 non-null   int64
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
```

```
[6]: df.isnull().sum()
```

```
[6]: Pregnancies      0
      Glucose         0
      BloodPressure    0
      SkinThickness    0
      Insulin          0
      BMI              0
      Pedigree         0
      Age              0
      Outcome          0
      dtype: int64
```

```
[8]: X = df.drop('Outcome', axis=1)
      y = df['Outcome']
```

```
[10]: from sklearn.model_selection import train_test_split
```

```
[11]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
      ↪ random_state=42)
```

1.2 Scaling the values

```
[13]: from sklearn.preprocessing import StandardScaler
```

```
[14]: scaler = StandardScaler()  
X_train = scaler.fit_transform(X_train)  
X_test = scaler.transform(X_test)
```

1.3 Training the Model

```
[15]: from sklearn.neighbors import KNeighborsClassifier
```

```
[16]: knn = KNeighborsClassifier(n_neighbors=5)  
knn.fit(X_train, y_train)
```

```
[16]: KNeighborsClassifier()
```

```
[17]: y_pred = knn.predict(X_test)
```

1.4 Evaluating Model

```
[26]: from sklearn.metrics import confusion_matrix, accuracy_score, precision_score,   
      ↪ recall_score, classification_report
```

```
[20]: cm = confusion_matrix(y_test, y_pred)  
print("Confusion Matrix:\n", cm)
```

```
Confusion Matrix:  
[[79 20]  
 [27 28]]
```

```
[22]: accuracy = accuracy_score(y_test, y_pred)  
print("Accuracy:", accuracy)
```

```
Accuracy: 0.6948051948051948
```

```
[23]: error_rate = 1 - accuracy  
print("Error Rate:", error_rate)
```

```
Error Rate: 0.30519480519480524
```

```
[24]: precision = precision_score(y_test, y_pred)  
print("Precision:", precision)
```

```
Precision: 0.5833333333333334
```

```
[25]: recall = recall_score(y_test, y_pred)  
print("Recall:", recall)
```

```
Recall: 0.509090909090909
```

1.5 Classification Report

```
[27]: print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
0	0.75	0.80	0.77	99
1	0.58	0.51	0.54	55
accuracy			0.69	154
macro avg	0.66	0.65	0.66	154
weighted avg	0.69	0.69	0.69	154

```
[ ]:
```