ML_Prac5

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0.0.1 Aim: Implement K-Nearest Neighbors algorithm on diabetes.csv dataset. Compute confusion matrix, accuracy, error rate, precision and recall on the given dataset.

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[1]: import pandas as pd
import numpy as np
from sklearn import metrics
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[2]: df = pd.read_csv("diabetes.csv")
df
```

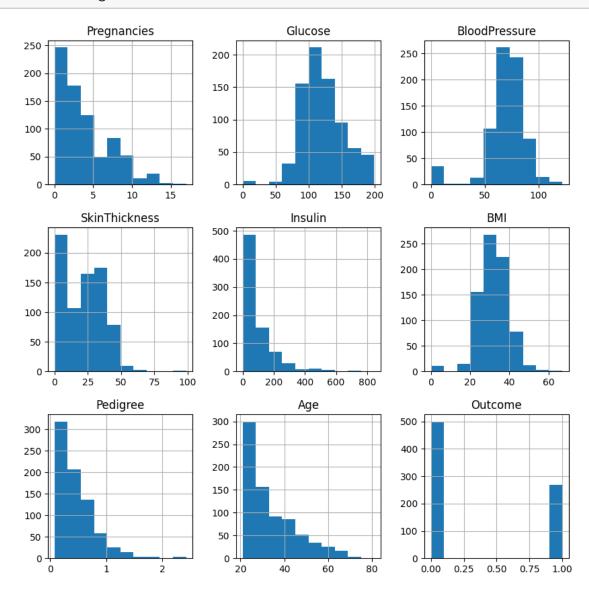
[2]:	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	
	•••	•••	•••		•••		
763	10	101	76	48	180	32.9	
764	2	122	70	27	0	36.8	
765	5	121	72	23	112	26.2	
766	1	126	60	0	0	30.1	
767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

	Pedigree	Age	Uutcome
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]
 [3]: df.shape
 [3]: (768, 9)
 [4]: # checking for null values
      df.isnull().any().value_counts()
 [4]: False
      Name: count, dtype: int64
 [5]: df.columns
 [5]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
             'BMI', 'Pedigree', 'Age', 'Outcome'],
            dtype='object')
 [6]: df_x = df.drop(columns='Outcome', axis=1)
      df_y = df['Outcome']
 [8]: print(df.isnull().sum())
     Pregnancies
                       0
     Glucose
                       0
     BloodPressure
                       0
     SkinThickness
                       0
     Insulin
                       0
     BMI
                       0
     Pedigree
                       0
                       0
     Age
     Outcome
                       0
     dtype: int64
 [9]: from sklearn.preprocessing import StandardScaler
      scale = StandardScaler()
      scaledX = scale.fit_transform(df_x)
[10]: # split into train and test
      from sklearn.model_selection import train_test_split
      x_train, x_test, y_train, y_test = train_test_split(scaledX, df_y, test_size=0.
       ⇒47, random_state=47)
```

[11]: # KNN from sklearn.neighbors import KNeighborsClassifier knn = KNeighborsClassifier(n_neighbors=7) knn.fit(x_train, y_train) y_pred = knn.predict(x_test)

[12]: p = df.hist(figsize = (10,10))



```
[13]: # Confusion matrix
cs = metrics.confusion_matrix(y_test,y_pred)
print("Confusion matrix: \n",cs)
```

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Confusion matrix:
      [[195 40]
      [ 72 54]]
[14]: # Accuracy score
      ac = metrics.accuracy_score(y_test, y_pred)
      print("Accuracy score: ",ac)
     Accuracy score: 0.6897506925207756
[15]: # Error rate (error_rate = 1- accuracy)
      er = 1-ac
      print("Error rate: ",er)
     Error rate: 0.3102493074792244
[16]: # Precision
      p = metrics.precision_score(y_test,y_pred)
      print("Precision: ", p)
     Precision: 0.574468085106383
[17]: # Recall
      r = metrics.recall_score(y_test,y_pred)
      print("Recall: ", r)
     Recall: 0.42857142857142855
[18]: # Classification report
      cr = metrics.classification_report(y_test,y_pred)
      print("Classification report: \n\n", cr)
     Classification report:
                    precision
                                 recall f1-score
                                                    support
                0
                        0.73
                                  0.83
                                            0.78
                                                       235
                        0.57
                                  0.43
                1
                                            0.49
                                                       126
                                                       361
                                            0.69
         accuracy
        macro avg
                        0.65
                                  0.63
                                            0.63
                                                       361
     weighted avg
                        0.68
                                  0.69
                                            0.68
                                                       361
```

[]: