

```

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

def calculate_accuracy(y_test, y_pred):
    correct = (y_pred == y_test).sum()
    total_instances = len(y_test)
    accuracy = correct / total_instances
    return correct, total_instances, accuracy

def euclidean_distance(x1, x2):
    distance = np.sqrt(np.sum((x1 - x2) ** 2))
    return distance

def count_zeros_ones(k_nearest, distances):
    count_0 = 0
    count_1 = 0
    for element in k_nearest:
        if element == 0:
            count_0 += 1
        elif element == 1:
            count_1 += 1

    if count_0 > count_1:
        return 0
    elif count_0 < count_1:
        return 1
    else:
        distances_0 = 0
        distances_1 = 0
        for i in range(len(k_nearest)):
            if k_nearest[i] == 0:
                distances_0 += 1/distances[i]
            elif k_nearest[i] == 1:
                distances_1 += 1/distances[i]
        if distances_0 > distances_1:
            return 0
        else:
            return 1

class KNN:
    def __init__(self):
        self.k = 1

    def setK(self, k):
        self.k = k

    def fit(self, X, y):

```

```

self.X_train = X
self.y_train = y

def predict(self, X_test):
    predictions = [self._predict(X_test[i]) for i in
range(X_test.shape[0])]
    return predictions

def _predict(self, x_test):
    distances = [euclidean_distance(x_test, x_train) for x_train
in self.X_train]
    # sort and return nums of rows before sort
    rows_num = np.argsort(distances)[:self.k]
    k_nearest = [self.y_train[i][0] for i in rows_num]
    distances = sorted(distances)
    predict_value = count_zeros_ones(k_nearest, distances)
    return predict_value

path = 'diabetes.csv'
diabetes = pd.read_csv(path)
print('diabetes')
print(diabetes)

```

```
diabetes
```

	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
..
762	9	89	62	0	0	22.5
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

	DiabetesPedigreeFunction	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
..
762	0.142	33	0
763	0.171	63	0
764	0.340	27	0
765	0.245	30	0
766	0.349	47	1

[767 rows x 9 columns]

the features and targets are separated

```
num_of_cols = diabetes.shape[1]
```

```
X = diabetes.iloc[:, 0:num_of_cols-1]
```

```
y = diabetes.iloc[:, num_of_cols-1:num_of_cols]
```

```
print("features")
```

```
print(X)
```

```
print("-----")
```

```
print("targets")
```

```
print(y)
```

features

	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
..
762	9	89	62	0	0	22.5
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

	DiabetesPedigreeFunction	Age
0	0.627	50
1	0.351	31
2	0.672	32
3	0.167	21
4	2.288	33
..
762	0.142	33
763	0.171	63
764	0.340	27
765	0.245	30
766	0.349	47

[767 rows x 8 columns]

	Outcome
0	1
1	0
2	1
3	0
4	1
..	...
762	0
763	0
764	0
765	0
766	1

[767 rows x 1 columns]

```
# divide data into 70% for training and 30% for testing
X_train, X_test, y_train, y_test = train_test_split(X, y,
shuffle=True, test_size=0.30, random_state=45)
print("X_train")
print(X_train)
print("-----")
print("X_test")
print(X_test)
print("-----")
print("y_train")
print(y_train)
print("-----")
print("y_test")
print(y_test)
```

X_train	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI
\						
172	2	87	0	23	0	28.9

405	2	123	48	32	165	42.1
357	13	129	0	30	0	39.9
483	0	84	82	31	125	38.2
737	8	65	72	23	0	32.0
..
725	4	112	78	40	0	39.4
607	1	92	62	25	41	19.5
544	1	88	78	29	76	32.0
643	4	90	0	0	0	28.0
414	0	138	60	35	167	34.6
DiabetesPedigreeFunction Age						
172		0.773	25			
405		0.520	26			
357		0.569	44			
483		0.233	23			
737		0.600	42			
..				
725		0.236	38			
607		0.482	25			
544		0.365	29			
643		0.610	31			
414		0.534	21			
[536 rows x 8 columns]						

X_test						
	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI
\						
195	5	158	84	41	210	39.4
51	1	101	50	15	36	24.2
66	0	109	88	30	0	32.5
437	5	147	75	0	0	29.9
664	6	115	60	39	0	33.7
..

38	2	90	68	42	0	38.2
354	3	90	78	0	0	42.7
18	1	103	30	38	83	43.3
229	0	117	80	31	53	45.2
540	8	100	74	40	215	39.4

	DiabetesPedigreeFunction	Age
195	0.395	29
51	0.526	26
66	0.855	38
437	0.434	28
664	0.245	40
..
38	0.503	27
354	0.559	21
18	0.183	33
229	0.089	24
540	0.661	43

[231 rows x 8 columns]

y_train

	Outcome
172	0
405	0
357	1
483	0
737	0
..	...
725	0
607	0
544	0
643	0
414	1

[536 rows x 1 columns]

y_test

	Outcome
195	1
51	0
66	1
437	0
664	1

```
..      ...
38      1
354     0
18      0
229     0
540     1
```

```
[231 rows x 1 columns]
```

```
# Normalize each feature column separately for training and test objects using Min-Max Scaling.
```

```
X_train = (X_train-X_train.min()) / (X_train.max()-X_train.min())
```

```
X_test = (X_test-X_test.min()) / (X_test.max()-X_test.min())
```

```
print("X_train after standardization")
```

```
print(X_train)
```

```
print("-----")
```

```
print("X_test after standardization")
```

```
print(X_test)
```

```
X_train after standardization
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin
BMI \					
172	0.117647	0.437186	0.000000	0.232323	0.000000
0.430700					
405	0.117647	0.618090	0.393443	0.323232	0.195035
0.627422					
357	0.764706	0.648241	0.000000	0.303030	0.000000
0.594635					
483	0.000000	0.422111	0.672131	0.313131	0.147754
0.569300					
737	0.470588	0.326633	0.590164	0.232323	0.000000
0.476900					
..
...					
725	0.235294	0.562814	0.639344	0.404040	0.000000
0.587183					
607	0.058824	0.462312	0.508197	0.252525	0.048463
0.290611					
544	0.058824	0.442211	0.639344	0.292929	0.089835
0.476900					
643	0.235294	0.452261	0.000000	0.000000	0.000000
0.417288					
414	0.000000	0.693467	0.491803	0.353535	0.197400
0.515648					

	DiabetesPedigreeFunction	Age
172	0.296755	0.078431
405	0.188728	0.098039

357	0.209650	0.450980
483	0.066183	0.039216
737	0.222886	0.411765
..
725	0.067464	0.333333
607	0.172502	0.078431
544	0.122545	0.156863
643	0.227156	0.196078
414	0.194705	0.000000

[536 rows x 8 columns]

X_test after standardization

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin
BMI \					
195	0.357143	0.802030	0.736842	0.788462	0.282258
0.687609					
51	0.071429	0.512690	0.438596	0.288462	0.048387
0.422339					
66	0.000000	0.553299	0.771930	0.576923	0.000000
0.567190					
437	0.357143	0.746193	0.657895	0.000000	0.000000
0.521815					
664	0.428571	0.583756	0.526316	0.750000	0.000000
0.588133					
..
...					
38	0.142857	0.456853	0.596491	0.807692	0.000000
0.666667					
354	0.214286	0.456853	0.684211	0.000000	0.000000
0.745201					
18	0.071429	0.522843	0.263158	0.730769	0.111559
0.755672					
229	0.000000	0.593909	0.701754	0.596154	0.071237
0.788831					
540	0.571429	0.507614	0.649123	0.769231	0.288978
0.687609					

	DiabetesPedigreeFunction	Age
195	0.138530	0.133333
51	0.196882	0.083333
66	0.343430	0.283333
437	0.155902	0.116667
664	0.071715	0.316667
..
38	0.186637	0.100000
354	0.211581	0.000000
18	0.044098	0.200000
229	0.002227	0.050000

540 0.257016 0.366667

[231 rows x 8 columns]

```
# Convert X to numpy array
```

```
X_train = X_train.to_numpy().reshape((-1, num_of_cols - 1))
```

```
X_test = X_test.to_numpy().reshape((-1, num_of_cols - 1))
```

```
# Convert y to numpy array
```

```
y_train = y_train.to_numpy()
```

```
y_test = y_test.to_numpy()
```

```
knn = KNN()
```

```
knn.fit(X_train, y_train)
```

```
avg = 0
```

```
iterations = 8
```

```
for k in range(2, iterations+2):
```

```
    print('k value:', k)
```

```
    knn.setK(k)
```

```
    y_pred = knn.predict(X_test)
```

```
    y_pred = np.array(y_pred).reshape(len(y_pred), 1)
```

```
    correct, total_instances, accuracy = calculate_accuracy(y_test,  
y_pred)
```

```
    avg += accuracy
```

```
    print('Number of correctly classified instances:', correct)
```

```
    print('Total number of instances:', total_instances)
```

```
    print('Accuracy:', accuracy)
```

```
    print()
```

k value: 2

Number of correctly classified instances: 155

Total number of instances: 231

Accuracy: 0.670995670995671

k value: 3

Number of correctly classified instances: 167

Total number of instances: 231

Accuracy: 0.7229437229437229

k value: 4

Number of correctly classified instances: 162

Total number of instances: 231

Accuracy: 0.7012987012987013

k value: 5

Number of correctly classified instances: 160

Total number of instances: 231

Accuracy: 0.6926406926406926

k value: 6

Number of correctly classified instances: 160

Total number of instances: 231
Accuracy: 0.6926406926406926

k value: 7
Number of correctly classified instances: 163
Total number of instances: 231
Accuracy: 0.7056277056277056

k value: 8
Number of correctly classified instances: 165
Total number of instances: 231
Accuracy: 0.7142857142857143

k value: 9
Number of correctly classified instances: 166
Total number of instances: 231
Accuracy: 0.7186147186147186

avg = avg/iterations
`print('The average accuracy across all iterations:', avg)`
The average accuracy across all iterations: 0.7023809523809524