CH-6 Classification Analysis Using Python

KNN(K Nearest Neighbor): -

```
In [14]: import pandas as pd
df = pd.read_csv("KNN.csv")
df
```

Out[14]:

	points	x1	x2	class
0	Α	1	2	0
1	В	2	3	0
2	С	3	1	1
3	D	6	5	1

```
In [15]: df = df.drop("points",axis=1)
df
```

Out[15]:

	x1	x2	class
0	1	2	0
1	2	3	0
2	3	1	1
3	6	5	1

Out[16]:

	^ 1	^	Ciass
0	1	2	0
1	2	3	0
2	3	1	1
3	6	5	1

Out[27]:

	x1	x2
3	6	5
1	2	3
0	1	2

```
In [37]:
         from sklearn.neighbors import KNeighborsClassifier
         nn = KNeighborsClassifier(n_neighbors=1)
         model = nn.fit(x_train,y_train)
In [38]: y_pred = model.predict(x_test)
In [39]: y_pred
Out[39]: array([0], dtype=int64)
In [40]: ## Confusion Metrics : -
         from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(y_test,y_pred)
         cm
Out[40]: array([[0, 0],
                [1, 0]], dtype=int64)
In [48]:
        TN = cm[0][0] # 0
         FP = cm[0][1] # 0
         FN = cm[1][0] # 1
         TP = cm[1][1] # 0
         print(TN)
         print(FP)
         print(FN)
         print(TP)
         0
         0
         1
         0
In [42]:
         accuracy = (TP+TN)/(TP+TN+FP+FN)
         accuracy
Out[42]: 0.0
In [43]:
         error_rate = (FP+FN)/(TP+TN+FP+FN)
         error_rate
Out[43]: 1.0
In [44]:
         error_rate = 1-accuracy
         error_rate
Out[44]: 1.0
In [45]:
         sensitivity = TP/(TP+FN)
         sensitivity
Out[45]: 0.0
```

```
In [46]: specificity = (TN)/(TN+FP)
specificity
```

<ipython-input-46-b01751f7ed08>:1: RuntimeWarning: invalid value encountered
in longlong_scalars
 specificity = (TN)/(TN+FP)

Out[46]: nan

Diabities

```
In [49]: import pandas as pd
    df = pd.read_csv("diabetes.csv")
    df
```

Out[49]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFuncti
0	6	148	72	35	0	33.6	0.6
1	1	85	66	29	0	26.6	0.3
2	8	183	64	0	0	23.3	0.6
3	1	89	66	23	94	28.1	0.1
4	0	137	40	35	168	43.1	2.2
763	10	101	76	48	180	32.9	0.1
764	2	122	70	27	0	36.8	0.3
765	5	121	72	23	112	26.2	0.2
766	1	126	60	0	0	30.1	0.3
767	1	93	70	31	0	30.4	0.3

768 rows × 9 columns

In [51]: y = df["Outcome"]
x = df.drop("Outcome",axis=1)
df

Out[51]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFuncti
0	6	148	72	35	0	33.6	0.6
1	1	85	66	29	0	26.6	0.3
2	8	183	64	0	0	23.3	0.6
3	1	89	66	23	94	28.1	0.1
4	0	137	40	35	168	43.1	2.2
763	10	101	76	48	180	32.9	0.1
764	2	122	70	27	0	36.8	0.3
765	5	121	72	23	112	26.2	0.2
766	1	126	60	0	0	30.1	0.3
767	1	93	70	31	0	30.4	0.3

768 rows × 9 columns

In [52]: from sklearn.model_selection import train_test_split
 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_stat
 x_train

Out[52]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFuncti
602	1	124	74	36	0	27.8	0.1
429	1	95	82	25	180	35.0	0.2
623	0	94	70	27	115	43.5	0.3
209	7	184	84	33	0	35.5	0.3
589	0	73	0	0	0	21.1	0.3
534	1	77	56	30	56	33.3	1.2
584	8	124	76	24	600	28.7	0.6
493	4	125	70	18	122	28.9	1.1
527	3	116	74	15	105	26.3	0.1
168	4	110	66	0	0	31.9	0.4

614 rows × 8 columns

```
In [63]:
        from sklearn.neighbors import KNeighborsClassifier
         nn = KNeighborsClassifier(n_neighbors=27)
         model = nn.fit(x_train,y_train)
In [64]: y_pred = model.predict(x_test)
In [65]: y_pred
Out[65]: array([0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
                0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0,
                0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0,
                1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0]
               dtype=int64)
         Confusion matrix
In [66]:
        from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(y_test,y_pred)
         cm
Out[66]: array([[98, 11],
                [24, 21]], dtype=int64)
In [67]:
        TN = cm[0][0]
         FP = cm[0][1]
         FN = cm[1][0]
         TP = cm[1][1]
         print(TN)
         print(FP)
         print(FN)
         print(TP)
         98
         11
         24
         21
In [68]:
         accuracy = (TP+TN)/(TP+TN+FP+FN)
         accuracy
Out[68]: 0.7727272727272727
In [69]:
         error_rate = (FP+FN)/(TP+TN+FP+FN)
         error_rate
Out[69]: 0.227272727272727
In [70]: |error_rate = 1-accuracy
         error_rate
Out[70]: 0.22727272727273
```

```
In [71]: sensitivity = TP/(TP+FN)
sensitivity
```

Out[71]: 0.466666666666667

```
In [62]: specificity = (TN)/(TN+FP)
specificity
```

Out[62]: 0.7431192660550459

Last Code

```
In [1]: import pandas as pd
    df = pd.read_csv("diabetes.csv")
    df
```

Out[1]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFuncti
0	6	148	72	35	0	33.6	0.6
1	1	85	66	29	0	26.6	0.3
2	8	183	64	0	0	23.3	0.6
3	1	89	66	23	94	28.1	0.1
4	0	137	40	35	168	43.1	2.2
763	10	101	76	48	180	32.9	0.1
764	2	122	70	27	0	36.8	0.3
765	5	121	72	23	112	26.2	0.2
766	1	126	60	0	0	30.1	0.3
767	1	93	70	31	0	30.4	0.3

768 rows × 9 columns

```
In [3]: y = df["Outcome"]
x = df.drop("Outcome",axis=1)
x
```

Out[3]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFuncti
0	6	148	72	35	0	33.6	0.6
1	1	85	66	29	0	26.6	0.3
2	8	183	64	0	0	23.3	0.6
3	1	89	66	23	94	28.1	0.1
4	0	137	40	35	168	43.1	2.2
763	10	101	76	48	180	32.9	0.1
764	2	122	70	27	0	36.8	0.3
765	5	121	72	23	112	26.2	0.2
766	1	126	60	0	0	30.1	0.3
767	1	93	70	31	0	30.4	0.3

768 rows × 8 columns

In [4]: from sklearn.model_selection import train_test_split
 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_stat
 x_train

Out[4]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFuncti
602	1	124	74	36	0	27.8	0.1
429	1	95	82	25	180	35.0	0.2
623	0	94	70	27	115	43.5	0.3
209	7	184	84	33	0	35.5	0.3
589	0	73	0	0	0	21.1	0.3
534	1	77	56	30	56	33.3	1.2
584	8	124	76	24	600	28.7	0.6
493	4	125	70	18	122	28.9	1.1
527	3	116	74	15	105	26.3	0.1
168	4	110	66	0	0	31.9	0.4

614 rows × 8 columns

In [18]: from sklearn.tree import DecisionTreeClassifier
nn = DecisionTreeClassifier(criterion="entropy", max_depth=27)

```
In [19]: |model = nn.fit(x_train,y_train)
In [20]: |y_pred = nn.predict(x_test)
         y_pred
Out[20]: array([0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1,
                0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0,
                0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0,
                0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
                0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0,
                0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0,
                1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0]
               dtype=int64)
In [21]: from sklearn.metrics import confusion_matrix
         cm = confusion matrix(y test,y pred)
         cm
Out[21]: array([[80, 29],
                [23, 22]], dtype=int64)
In [22]: |TN = cm[0][0]
         FP = cm[0][1]
         FN = cm[1][0]
         TP = cm[1][1]
         print(TN)
         print(FP)
         print(FN)
         print(TP)
         80
         29
         23
         22
         accuracy = (TP+TN)/(TP+TN+FP+FN)
In [24]:
         accuracy
Out[24]: 0.6623376623376623
In [25]: error_rate = (FP+FN)/(TP+TN+FP+FN)
         error rate
Out[25]: 0.33766233766233766
In [26]:
         error_rate = 1-accuracy
         error_rate
Out[26]: 0.33766233766233766
         sensitivity = TP/(TP+FN)
In [27]:
         sensitivity
Out[27]: 0.488888888888888
```

```
In [28]:
         specificity = (TN)/(TN+FP)
         specificity
Out[28]: 0.7339449541284404
In [30]: | from sklearn import tree
         gr = tree.export_text(nn)
         print(gr)
                                              |--- class: 0
                                      --- feature_7 > 39.00
                                         |--- class: 1
                                   --- feature_6 > 0.31
                                      |--- feature_5 <= 45.30
                                         |--- class: 1
                                      --- feature_5 > 45.30
                                         |--- class: 0
                              --- feature_7 > 66.50
                                 |--- class: 0
                         |--- feature 3 > 32.50
                            |--- class: 1
                      --- feature_6 > 2.23
                        |--- class: 0
                  --- feature_2 > 93.00
                     |--- feature_0 <= 7.50
                         |--- feature_0 <= 2.00
                             |--- class: 1
                          --- feature_0 > 2.00
In [31]:
         pre = [[10,20,30,40,50,60,70,80]]
         test_data =nn.predict(pre)
         test_data
Out[31]: array([0], dtype=int64)
 In [ ]:
```