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Program: MCA-I

Subject: AI-ML

Question 2

GitHub: https://github.com/navYadav20/AI-ML

KNN classification

```
In [2]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.metrics import accuracy_score
```

In [3]: df = pd.read_csv('D:/ai-ml assignment/assignment 4/cancer.csv')
 df.head(5)

Out[3]:

	id	clump_thickness	unif_cell_size	unif_cell_shape	marg_adhesion	single_epith_cel
-	1000025	5	1	1	1	
1	1002945	5	4	4	5	
2	1015425	3	1	1	1	
3	1016277	6	8	8	1	
4	1017023	4	1	1	3	
4						•

```
In [4]:
    df.replace('?', -99999, inplace=True)
    df.drop(columns=['id'], inplace=True)
    df
```

Out[4]:

	clump_thickness	unif_cell_size	unif_cell_shape	marg_adhesion	single_epith_cell_size
0	5	1	1	1	2
1	5	4	4	5	7
2	3	1	1	1	2
3	6	8	8	1	3
4	4	1	1	3	2
694	3	1	1	1	3
695	2	1	1	1	2
696	5	10	10	3	7
697	4	8	6	4	3
698	4	8	8	5	4

699 rows × 10 columns

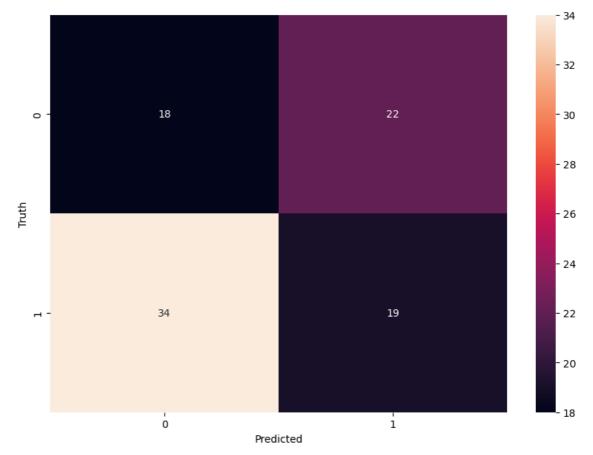
```
In [5]: X = np.array(df.drop('classes', axis=1)) # Dropping 'classes' column to cr
X
```

```
y = np.array(df['classes']) # Assigning the 'classes' column to the target
In [6]:
Out[6]: array([0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
              0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1,
              1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1,
              0, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1,
              0, 0, 0, 0, 0, 0, 0, 0, 0, 1,
                                            1, 1, 1, 0, 1,
                                                          1,
                                                             1, 1, 1, 0, 1,
              0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1,
                                                            0, 1, 0,
              1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0,
              0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1,
                                                             0, 0, 0, 0, 0,
              0, 1, 0,
                            0, 0, 1, 1, 0, 1,
                                            1, 1, 0, 1, 1,
                      1,
                         0,
                                                          0,
              0, 0, 1, 1, 0, 0, 0, 1, 1,
                                      0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0,
              0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0,
              0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1,
                0, 1,
                      1,
                         1, 0, 1, 0, 1, 1,
                                         0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1,
              1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0,
              1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1,
              1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1,
                                            0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0,
              0, 1, 0, 0, 1, 1,
                              1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0,
                                                             0, 0, 0,
              0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
              0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0,
              0, 0, 0,
                      1,
                         0, 0, 0, 1, 0, 1, 0,
                                            0, 0, 0, 0, 0,
                                                          0,
                                                             1, 1, 0, 0, 0,
                0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1,
                                                            1, 0, 0, 0, 0,
              0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1,
              0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                   0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0,
                                                          1,
                                                             1, 0, 0, 0, 0,
              0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0,
                                                          1, 0, 1, 1, 1, 0,
              1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0,
              0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0,
              0, 0, 0, 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, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1], dtype=int64)

```
In [7]: import seaborn as sns

cm = [[18, 22], [ 34, 19]]
  plt.figure(figsize=(10,7))
  sns.heatmap(cm, annot=True, fmt='d')
  plt.xlabel('Predicted')
  plt.ylabel('Truth')
  plt.show()
```



```
In [8]: # Splitting the dataset into the Training set and Test set
    from sklearn.model_selection import train_test_split
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.35,
```

```
In [9]: # Feature Scaling
    from sklearn.preprocessing import StandardScaler
    sc = StandardScaler()
    X_train = sc.fit_transform(X_train)
    X_test = sc.transform(X_test)
```

```
In [10]: #principle component analysis

from sklearn.decomposition import PCA
pca = PCA(n_components=2)
X_train = pca.fit_transform(X_train)
X_test = pca.fit_transform(X_test)
explained_variance=pca.explained_variance_ratio_
```

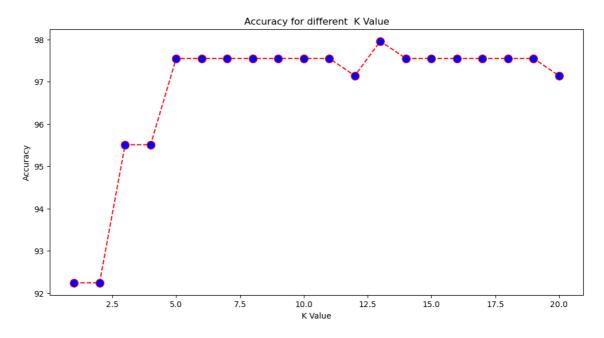
```
from sklearn.neighbors import KNeighborsClassifier
In [11]:
         knn = []
         for i in range(1,21):
             classifier = KNeighborsClassifier(n neighbors=i)
             trained_model=classifier.fit(X_train,y_train)
             trained_model.fit(X_train,y_train )
             # Predicting the Test set results
             y_pred = classifier.predict(X_test)
             # Making the Confusion Matrix
             from sklearn.metrics import confusion_matrix
             cm_KNN = confusion_matrix(y_test, y_pred)
             print(cm_KNN)
             print("Accuracy score of train KNN")
             print(accuracy_score(y_train, trained_model.predict(X_train))*100)
             print("Accuracy score of test KNN")
             print(accuracy_score(y_test, y_pred)*100)
             knn.append(accuracy_score(y_test, y_pred)*100)
```

```
[[154 10]
[ 9 72]]
Accuracy score of train KNN
100.0
Accuracy score of test KNN
92.24489795918367
[[160 4]
[ 15 66]]
Accuracy score of train KNN
97.79735682819384
Accuracy score of test KNN
92.24489795918367
[[160 4]
[ 7 74]]
Accuracy score of train KNN
96.91629955947137
Accuracy score of test KNN
95.51020408163265
[[160
      4]
[ 7 74]]
Accuracy score of train KNN
96.47577092511013
Accuracy score of test KNN
95.51020408163265
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.69603524229075
Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.69603524229075
Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 2 79]]
Accuracy score of train KNN
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Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.91629955947137
Accuracy score of test KNN
97.55102040816327
[[160
      4]
[ 2 79]]
Accuracy score of train KNN
96.69603524229075
Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.69603524229075
Accuracy score of test KNN
97.55102040816327
[[160
       4]
```

```
[ 2 79]]
Accuracy score of train KNN
96.69603524229075
Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 3 78]]
Accuracy score of train KNN
96.69603524229075
Accuracy score of test KNN
97.14285714285714
[[160 4]
[ 1 80]]
Accuracy score of train KNN
96.25550660792952
Accuracy score of test KNN
97.95918367346938
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.25550660792952
Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.25550660792952
Accuracy score of test KNN
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      4]
[ 2 79]]
Accuracy score of train KNN
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Accuracy score of test KNN
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Accuracy score of train KNN
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Accuracy score of test KNN
97.55102040816327
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Accuracy score of train KNN
96.25550660792952
Accuracy score of test KNN
97.55102040816327
[[160 4]
[ 2 79]]
Accuracy score of train KNN
96.25550660792952
Accuracy score of test KNN
97.55102040816327
[[160
      4]
[ 3 78]]
Accuracy score of train KNN
96.0352422907489
Accuracy score of test KNN
97.14285714285714
```

```
In [12]: plt.figure(figsize=(12, 6))
    plt.plot(range(1, 21),knn, color='red', linestyle='dashed', marker='o', mar
    plt.title('Accuracy for different K Value')
    plt.xlabel('K Value')
    plt.ylabel('Accuracy')
```

Out[12]: Text(0, 0.5, 'Accuracy')



```
In [13]: # Fitting SVM to the Training set

from sklearn.svm import SVC
classifier = SVC(kernel = 'linear', random_state = 0)

trained_model=classifier.fit(X_train,y_train)
trained_model.fit(X_train,y_train)
```

```
Out[13]: SVC SVC(kernel='linear', random_state=0)
```

```
In [14]: # Predicting the Test set results

y_pred = classifier.predict(X_test)
```

```
In [15]: # Making the Confusion Matrix
         from sklearn.metrics import confusion_matrix
         cm_SVM = confusion_matrix(y_test, y_pred)
         print(cm SVM)
         print("Accuracy score of train SVM")
         print(accuracy_score(y_train, trained_model.predict(X_train))*100)
         [[160
                 4]
          [ 4 77]]
         Accuracy score of train SVM
         96.47577092511013
In [16]: print("Accuracy score of test SVM")
         print(accuracy_score(y_test, y_pred)*100)
         Accuracy score of test SVM
         96.73469387755102
In [17]: from sklearn.preprocessing import StandardScaler
         X_train = sc.fit_transform(X_train)
         X_test = sc.transform(X_test)
         # Example query
         query_data = pd.DataFrame({
             'unif_cell_size': [2],
             'unif_cell_shape': [3]
         })
         # Scale the query data using the same scaler
         query_data_scaled = sc.transform(query_data)
         # Make predictions for the query data
         query_prediction = classifier.predict(query_data_scaled)
         print(f"Predicted class for query data: {query_prediction}")
         Predicted class for query data: [0]
         C:\Users\msuse\anaconda3\Lib\site-packages\sklearn\base.py:457: UserWarnin
         g: X has feature names, but StandardScaler was fitted without feature name
         S
           warnings.warn(
 In [ ]:
```