

Classification

```
1)KNN:  
    -k Nearest  
    -k Neighbors  
  
2)Desicion Tree:
```

KNN algorithm

```
In [27]: import pandas as pd  
df=pd.read_csv('tshirt.csv')  
df
```

Out[27]:

| | Height | Width | Size |
|----|--------|-------|------|
| 0 | 158 | 58 | M |
| 1 | 158 | 59 | M |
| 2 | 158 | 63 | M |
| 3 | 160 | 59 | M |
| 4 | 160 | 60 | M |
| 5 | 163 | 60 | M |
| 6 | 163 | 61 | M |
| 7 | 160 | 64 | L |
| 8 | 163 | 64 | L |
| 9 | 165 | 61 | L |
| 10 | 165 | 62 | L |
| 11 | 165 | 65 | L |
| 12 | 168 | 62 | L |
| 13 | 168 | 63 | L |
| 14 | 168 | 66 | L |
| 15 | 170 | 63 | L |
| 16 | 170 | 64 | L |
| 17 | 170 | 68 | L |

In [28]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18 entries, 0 to 17
Data columns (total 3 columns):
 #   Column  Non-Null Count  Dtype  
---  -
 0   Height  18 non-null      int64  
 1   Width   18 non-null      int64  
 2   Size    18 non-null      object  
dtypes: int64(2), object(1)
memory usage: 560.0+ bytes
```

In [29]: `df=pd.get_dummies(df,drop_first=True)`
`df`

Out[29]:

| | Height | Width | Size_M |
|----|--------|-------|--------|
| 0 | 158 | 58 | 1 |
| 1 | 158 | 59 | 1 |
| 2 | 158 | 63 | 1 |
| 3 | 160 | 59 | 1 |
| 4 | 160 | 60 | 1 |
| 5 | 163 | 60 | 1 |
| 6 | 163 | 61 | 1 |
| 7 | 160 | 64 | 0 |
| 8 | 163 | 64 | 0 |
| 9 | 165 | 61 | 0 |
| 10 | 165 | 62 | 0 |
| 11 | 165 | 65 | 0 |
| 12 | 168 | 62 | 0 |
| 13 | 168 | 63 | 0 |
| 14 | 168 | 66 | 0 |
| 15 | 170 | 63 | 0 |
| 16 | 170 | 64 | 0 |
| 17 | 170 | 68 | 0 |

```
In [30]: y=df['Size_M']  
y
```

```
Out[30]: 0      1  
1      1  
2      1  
3      1  
4      1  
5      1  
6      1  
7      0  
8      0  
9      0  
10     0  
11     0  
12     0  
13     0  
14     0  
15     0  
16     0  
17     0  
Name: Size_M, dtype: uint8
```

```
In [31]: x=df.drop('Size_M',axis=1)  
x
```

```
Out[31]:
```

| | Height | Width |
|----|--------|-------|
| 0 | 158 | 58 |
| 1 | 158 | 59 |
| 2 | 158 | 63 |
| 3 | 160 | 59 |
| 4 | 160 | 60 |
| 5 | 163 | 60 |
| 6 | 163 | 61 |
| 7 | 160 | 64 |
| 8 | 163 | 64 |
| 9 | 165 | 61 |
| 10 | 165 | 62 |
| 11 | 165 | 65 |
| 12 | 168 | 62 |
| 13 | 168 | 63 |
| 14 | 168 | 66 |
| 15 | 170 | 63 |
| 16 | 170 | 64 |
| 17 | 170 | 68 |

```
In [32]: from sklearn.neighbors import KNeighborsClassifier
```

```
In [64]: nn=KNeighborsClassifier(n_neighbors=3)
model=nn.fit(x,y)
y_pred=model.predict(x)
print(y_pred)
```

```
[1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0]
```

```
In [65]: y_pred=model.predict([[161,61]])
print(y_pred)
```

```
[1]
```

```
In [66]: y_pred=model.predict(x)
print(y_pred)
dif=pd.DataFrame({'Actual':y, 'Predicated':y_pred})
dif
```

```
[1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0]
```

Out[66]:

| | Actual | Predicated |
|----|--------|------------|
| 0 | 1 | 1 |
| 1 | 1 | 1 |
| 2 | 1 | 1 |
| 3 | 1 | 1 |
| 4 | 1 | 1 |
| 5 | 1 | 1 |
| 6 | 1 | 1 |
| 7 | 0 | 0 |
| 8 | 0 | 0 |
| 9 | 0 | 0 |
| 10 | 0 | 0 |
| 11 | 0 | 0 |
| 12 | 0 | 0 |
| 13 | 0 | 0 |
| 14 | 0 | 0 |
| 15 | 0 | 0 |
| 16 | 0 | 0 |
| 17 | 0 | 0 |

```
In [67]: from sklearn.metrics import accuracy_score
acs=accuracy_score(y,y_pred)
print("acs :", acs)
```

acs : 1.0

```
In [68]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y,y_pred)
cm
```

```
Out[68]: array([[11,  0],
               [ 0,  7]], dtype=int64)
```

```
In [69]: from sklearn.metrics import accuracy_score
print("error :",1-ac)
```

error : 0.0

accuracy = $TP+TN / TP+TN+FP+FN$

error = $FP+FN / TP+TN+FP+FN$ or $1-\text{accuracy}$

sensitivity = $TP / TP+FN$

specificity = $TN / TN+FP$

```
In [70]: TN=cm[0][0]
FP=cm[0][1]
FN=cm[1][0]
TP=cm[1][1]
```

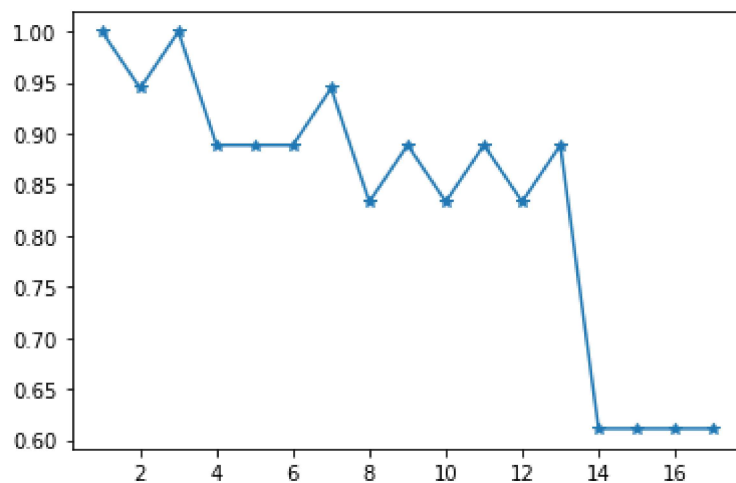
```
In [71]: print("sensitivity :",TP / (TP+FN))
```

sensitivity : 1.0

```
In [72]: print("specificity :",TN / (TN+FP))
```

specificity : 1.0

```
In [75]: import matplotlib.pyplot as plt
k=[]
for i in range(1,18):
    nn=KNeighborsClassifier(n_neighbors=i)
    model=nn.fit(x,y)
    y_pred=model.predict(x)
    # print(y_pred)
    acc=accuracy_score(y,y_pred)
    k.append(acc)
plt.plot(range(1,18),k,marker='*')
plt.show()
```



KNN Model-1

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

Out[1]:

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | DiabetesPedigreeFunction |
|-----|-------------|---------|---------------|---------------|---------|------|--------------------------|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.62 |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.35 |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.67 |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.16 |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.28 |
| ... | ... | ... | ... | ... | ... | ... | . |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.17 |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.34 |
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.24 |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.34 |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.31 |

768 rows × 9 columns



```
In [2]: 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   DiabetesPedigreeFunction 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
```

```
In [3]: y=df['Outcome']
y
```

```
Out[3]: 0      1
1      0
2      1
3      0
4      1
      ..
763    0
764    0
765    0
766    1
767    0
Name: Outcome, Length: 768, dtype: int64
```

```
In [4]: x=df.drop("Outcome",axis=1)
x
```

```
Out[4]:
```

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | DiabetesPedigreeFuncio |
|-----|-------------|---------|---------------|---------------|---------|------|------------------------|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.62 |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.35 |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.67 |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.16 |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.28 |
| ... | ... | ... | ... | ... | ... | ... | . |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.17 |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.34 |
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.24 |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.34 |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.31 |

768 rows × 8 columns



```
In [6]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.8,random_state=
```

```
In [9]: print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(614, 8)
(154, 8)
(614,)
(154,)
```



```
In [102]: from sklearn.neighbors import KNeighborsClassifier
nn=KNeighborsClassifier(n_neighbors=15)
model=nn.fit(x_train,y_train)
y_pred=model.predict(x_test)
print(y_pred)
```

```
[1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 0 0 0 0 1 1 0 1 1 1 0 0 0 0 0 1 0 1 0
 0 0 0 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 0 1 0 0 0 1 0 1 0 0 0 0 1 0 1 1 1 1 0
 1 0 1 0 0 1 0 0 0 0 0 1 1 0 1 0 0 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 1 0 0 0 1 0 0 0 0 0 0 1 1 1 0 1 0 0 0 0 0 0 0 1 0 0 1 0 1 0 0
 0 0 0 1 0 0]
```

```
In [103]: dif=pd.DataFrame({'Actual':y_test,'Predicted':y_pred})
dif
```

Out[103]:

| | Actual | Predicted |
|-----|--------|-----------|
| 285 | 0 | 1 |
| 101 | 0 | 0 |
| 581 | 0 | 0 |
| 352 | 0 | 0 |
| 726 | 0 | 0 |
| ... | ... | ... |
| 563 | 0 | 0 |
| 318 | 0 | 0 |
| 154 | 1 | 1 |
| 684 | 0 | 0 |
| 643 | 0 | 0 |

154 rows × 2 columns

```
In [104]: dif.to_csv("output.csv")
```

```
In [105]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
cm
```

Out[105]: array([[89, 10],
[23, 32]], dtype=int64)

```
In [106]: TN=cm[0][0]
FP=cm[0][1]
FN=cm[1][0]
TP=cm[1][1]
```

```
In [107]: from sklearn.metrics import accuracy_score
acs=accuracy_score(y_test,y_pred)
print("acs :", acs)
```

acs : 0.7857142857142857

```
In [108]: print("error :",1-ac)
```

error : 0.2142857142857143

```
In [109]: print("sensitivity :",TP / (TP+FN))
```

sensitivity : 0.5818181818181818

```
In [110]: print("specificity :",TN / (TN+FP))
```

specificity : 0.898989898989899

```
In [111]: import matplotlib.pyplot as plt
k=[]
for i in range(1,155):
    nn=KNeighborsClassifier(n_neighbors=i)
    model=nn.fit(x_train,y_train)
    y_pred=model.predict(x_test)
    # print(y_pred)
    acc=accuracy_score(y_test,y_pred)
    k.append(acc)
    print(i,':',acc)
plt.plot(range(1,155),k,marker='*')
plt.show()
```

1 : 0.7207792207792207
2 : 0.7272727272727273
3 : 0.7402597402597403
4 : 0.7207792207792207
5 : 0.7337662337662337
6 : 0.7727272727272727
7 : 0.7597402597402597
8 : 0.7662337662337663
9 : 0.7662337662337663
10 : 0.7597402597402597
11 : 0.7467532467532467
12 : 0.7727272727272727
13 : 0.7662337662337663
14 : 0.7727272727272727
15 : 0.7857142857142857
16 : 0.7857142857142857
17 : 0.7857142857142857
18 : 0.7727272727272727
19 : 0.7727272727272727
20 : 0.7727272727272727

KNN Model-2

```
In [120]: import pandas as pd
df=pd.read_csv('AptitudeCommunication.csv')
df
```

Out[120]:

| | Name | Aptitude | Communication | Class |
|----|----------|----------|---------------|---------|
| 0 | Karuna | 2 | 5.0 | Speaker |
| 1 | Bhavan | 2 | 6.0 | Speaker |
| 2 | Gaurav | 7 | 6.0 | Leader |
| 3 | Parul | 7 | 5.0 | Leader |
| 4 | Dinesh | 8 | 6.0 | Leader |
| 5 | Jani | 4 | 7.0 | Speaker |
| 6 | Bobby | 6 | 6.0 | Leader |
| 7 | Parimal | 3 | 5.5 | Speaker |
| 8 | Govind | 8 | 6.0 | Leader |
| 9 | Shushant | 6 | 4.0 | Leader |
| 10 | Gauri | 6 | 5.5 | Speaker |
| 11 | Bharat | 6 | 7.0 | Leader |
| 12 | Rajvi | 6 | 6.0 | Speaker |
| 13 | Pradeep | 9 | 7.0 | Leader |

```
In [121]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14 entries, 0 to 13
Data columns (total 4 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Name            14 non-null    object
1   Aptitude        14 non-null    int64
2   Communication    14 non-null    float64
3   Class           14 non-null    object
dtypes: float64(1), int64(1), object(2)
memory usage: 576.0+ bytes
```

```
In [122]: df.drop(["Name"],axis=1,inplace=True)
df
```

Out[122]:

| | Aptitude | Communication | Class |
|----|----------|---------------|---------|
| 0 | 2 | 5.0 | Speaker |
| 1 | 2 | 6.0 | Speaker |
| 2 | 7 | 6.0 | Leader |
| 3 | 7 | 5.0 | Leader |
| 4 | 8 | 6.0 | Leader |
| 5 | 4 | 7.0 | Speaker |
| 6 | 6 | 6.0 | Leader |
| 7 | 3 | 5.5 | Speaker |
| 8 | 8 | 6.0 | Leader |
| 9 | 6 | 4.0 | Leader |
| 10 | 6 | 5.5 | Speaker |
| 11 | 6 | 7.0 | Leader |
| 12 | 6 | 6.0 | Speaker |
| 13 | 9 | 7.0 | Leader |

```
In [125]: df1=pd.get_dummies(df,drop_first=True)
df1
```

Out[125]:

| | Aptitude | Communication | Class_Speaker |
|----|----------|---------------|---------------|
| 0 | 2 | 5.0 | 1 |
| 1 | 2 | 6.0 | 1 |
| 2 | 7 | 6.0 | 0 |
| 3 | 7 | 5.0 | 0 |
| 4 | 8 | 6.0 | 0 |
| 5 | 4 | 7.0 | 1 |
| 6 | 6 | 6.0 | 0 |
| 7 | 3 | 5.5 | 1 |
| 8 | 8 | 6.0 | 0 |
| 9 | 6 | 4.0 | 0 |
| 10 | 6 | 5.5 | 1 |
| 11 | 6 | 7.0 | 0 |
| 12 | 6 | 6.0 | 1 |
| 13 | 9 | 7.0 | 0 |

```
In [126]: print(type(df1))
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
In [127]: y=df1['Class_Speaker']  
y
```

```
Out[127]: 0      1  
1      1  
2      0  
3      0  
4      0  
5      1  
6      0  
7      1  
8      0  
9      0  
10     1  
11     0  
12     1  
13     0  
Name: Class_Speaker, dtype: uint8
```

```
In [128]: x=df1.drop("Class_Speaker",axis=1)  
x
```

```
Out[128]:
```

| | Aptitude | Communication |
|----|----------|---------------|
| 0 | 2 | 5.0 |
| 1 | 2 | 6.0 |
| 2 | 7 | 6.0 |
| 3 | 7 | 5.0 |
| 4 | 8 | 6.0 |
| 5 | 4 | 7.0 |
| 6 | 6 | 6.0 |
| 7 | 3 | 5.5 |
| 8 | 8 | 6.0 |
| 9 | 6 | 4.0 |
| 10 | 6 | 5.5 |
| 11 | 6 | 7.0 |
| 12 | 6 | 6.0 |
| 13 | 9 | 7.0 |

```
In [143]: from sklearn.neighbors import KNeighborsClassifier
nn=KNeighborsClassifier(n_neighbors=3)
model=nn.fit(x,y)
y_pred=model.predict(x)
print(y_pred)
```

```
[1 1 0 0 0 1 1 1 0 0 1 0 1 0]
```

```
In [144]: dif=pd.DataFrame({'Actual':y, 'Predicted':y_pred})
dif
```

Out[144]:

| | Actual | Predicted |
|----|--------|-----------|
| 0 | 1 | 1 |
| 1 | 1 | 1 |
| 2 | 0 | 0 |
| 3 | 0 | 0 |
| 4 | 0 | 0 |
| 5 | 1 | 1 |
| 6 | 0 | 1 |
| 7 | 1 | 1 |
| 8 | 0 | 0 |
| 9 | 0 | 0 |
| 10 | 1 | 1 |
| 11 | 0 | 0 |
| 12 | 1 | 1 |
| 13 | 0 | 0 |

```
In [145]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y,y_pred)
cm
```

Out[145]: array([[7, 1],
[0, 6]], dtype=int64)

```
In [146]: TN=cm[0][0]
FP=cm[0][1]
FN=cm[1][0]
TP=cm[1][1]
```

```
In [147]: from sklearn.metrics import accuracy_score
acs=accuracy_score(y,y_pred)
print("acs :",acs)
```

```
acs : 0.9285714285714286
```

```
In [148]: print("error :",1-acs)
```

```
error : 0.0714285714285714
```

```
In [149]: print("sensitivity :",TP / (TP+FN))
```

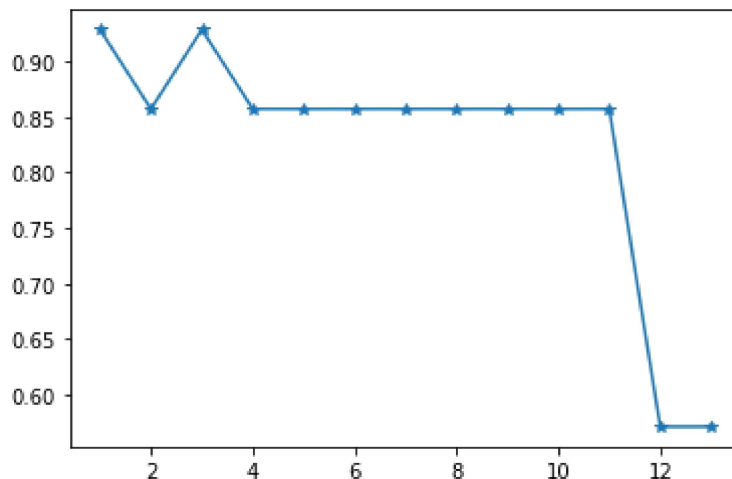
```
sensitivity : 1.0
```

```
In [150]: print("specificity :",TN / (TN+FP))
```

```
specificity : 0.875
```

```
In [151]: import matplotlib.pyplot as plt
k=[]
for i in range(1,14):
    nn=KNeighborsClassifier(n_neighbors=i)
    model=nn.fit(x,y)
    y_pred=model.predict(x)
    # print(y_pred)
    acc=accuracy_score(y,y_pred)
    k.append(acc)
    print(i,':',acc)
plt.plot(range(1,14),k,marker='*')
plt.show()
```

```
1 : 0.9285714285714286
2 : 0.8571428571428571
3 : 0.9285714285714286
4 : 0.8571428571428571
5 : 0.8571428571428571
6 : 0.8571428571428571
7 : 0.8571428571428571
8 : 0.8571428571428571
9 : 0.8571428571428571
10 : 0.8571428571428571
11 : 0.8571428571428571
12 : 0.5714285714285714
13 : 0.5714285714285714
```



Decision Tree

Tree-1


```
In [3]: import pandas as pd
df=pd.read_csv('DecisionTree_Sports.csv')
df
```

Out[3]:

| | Day | Weather | Temp | Humidity | Wind | Play_Sports |
|----|-------|---------|------|----------|--------|-------------|
| 0 | Day1 | Sunny | Hot | High | Weak | No |
| 1 | Day2 | Sunny | Hot | High | Strong | No |
| 2 | Day3 | Cloudy | Hot | High | Weak | Yes |
| 3 | Day4 | Rainy | Mild | High | Weak | Yes |
| 4 | Day5 | Rainy | Cool | Normal | Weak | Yes |
| 5 | Day6 | Rainy | Cool | Normal | Strong | No |
| 6 | Day7 | Cloudy | Cool | Normal | Strong | Yes |
| 7 | Day8 | Sunny | Mild | High | Weak | No |
| 8 | Day9 | Sunny | Cool | Normal | Weak | Yes |
| 9 | Day10 | Rainy | Mild | Normal | Weak | Yes |
| 10 | Day11 | Sunny | Mild | Normal | Strong | Yes |
| 11 | Day12 | Cloudy | Mild | High | Strong | Yes |
| 12 | Day13 | Cloudy | Hot | Normal | Weak | Yes |
| 13 | Day14 | Rainy | Mild | High | Strong | No |

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14 entries, 0 to 13
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Day              14 non-null    object
1   Weather          14 non-null    object
2   Temp             14 non-null    object
3   Humidity         14 non-null    object
4   Wind             14 non-null    object
5   Play_Sports      14 non-null    object
dtypes: object(6)
memory usage: 800.0+ bytes
```

```
In [5]: df.drop("Day",axis=1,inplace=True)
df
```

Out[5]:

| | Weather | Temp | Humidity | Wind | Play_Sports |
|----|---------|------|----------|--------|-------------|
| 0 | Sunny | Hot | High | Weak | No |
| 1 | Sunny | Hot | High | Strong | No |
| 2 | Cloudy | Hot | High | Weak | Yes |
| 3 | Rainy | Mild | High | Weak | Yes |
| 4 | Rainy | Cool | Normal | Weak | Yes |
| 5 | Rainy | Cool | Normal | Strong | No |
| 6 | Cloudy | Cool | Normal | Strong | Yes |
| 7 | Sunny | Mild | High | Weak | No |
| 8 | Sunny | Cool | Normal | Weak | Yes |
| 9 | Rainy | Mild | Normal | Weak | Yes |
| 10 | Sunny | Mild | Normal | Strong | Yes |
| 11 | Cloudy | Mild | High | Strong | Yes |
| 12 | Cloudy | Hot | Normal | Weak | Yes |
| 13 | Rainy | Mild | High | Strong | No |

```
In [6]: df1=pd.get_dummies(df,drop_first=True)
df1
```

Out[6]:

| | Weather_Rainy | Weather_Sunny | Temp_Hot | Temp_Mild | Humidity_Normal | Wind_Weak | Play_S |
|----|---------------|---------------|----------|-----------|-----------------|-----------|--------|
| 0 | 0 | 1 | 1 | 0 | 0 | 1 | |
| 1 | 0 | 1 | 1 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 1 | 0 | 0 | 1 | |
| 3 | 1 | 0 | 0 | 1 | 0 | 1 | |
| 4 | 1 | 0 | 0 | 0 | 1 | 1 | |
| 5 | 1 | 0 | 0 | 0 | 1 | 0 | |
| 6 | 0 | 0 | 0 | 0 | 1 | 0 | |
| 7 | 0 | 1 | 0 | 1 | 0 | 1 | |
| 8 | 0 | 1 | 0 | 0 | 1 | 1 | |
| 9 | 1 | 0 | 0 | 1 | 1 | 1 | |
| 10 | 0 | 1 | 0 | 1 | 1 | 0 | |
| 11 | 0 | 0 | 0 | 1 | 0 | 0 | |
| 12 | 0 | 0 | 1 | 0 | 1 | 1 | |
| 13 | 1 | 0 | 0 | 1 | 0 | 0 | |

```
In [7]: y=df1['Play_Sports_Yes']
y
```

```
Out[7]: 0      0
1      0
2      1
3      1
4      1
5      0
6      1
7      0
8      1
9      1
10     1
11     1
12     1
13     0
Name: Play_Sports_Yes, dtype: uint8
```

```
In [8]: x=df1.drop("Play_Sports_Yes",axis=1)
x
```

```
Out[8]:
```

| | Weather_Rainy | Weather_Sunny | Temp_Hot | Temp_Mild | Humidity_Normal | Wind_Weak |
|----|---------------|---------------|----------|-----------|-----------------|-----------|
| 0 | 0 | 1 | 1 | 0 | 0 | 1 |
| 1 | 0 | 1 | 1 | 0 | 0 | 0 |
| 2 | 0 | 0 | 1 | 0 | 0 | 1 |
| 3 | 1 | 0 | 0 | 1 | 0 | 1 |
| 4 | 1 | 0 | 0 | 0 | 1 | 1 |
| 5 | 1 | 0 | 0 | 0 | 1 | 0 |
| 6 | 0 | 0 | 0 | 0 | 1 | 0 |
| 7 | 0 | 1 | 0 | 1 | 0 | 1 |
| 8 | 0 | 1 | 0 | 0 | 1 | 1 |
| 9 | 1 | 0 | 0 | 1 | 1 | 1 |
| 10 | 0 | 1 | 0 | 1 | 1 | 0 |
| 11 | 0 | 0 | 0 | 1 | 0 | 0 |
| 12 | 0 | 0 | 1 | 0 | 1 | 1 |
| 13 | 1 | 0 | 0 | 1 | 0 | 0 |

```
In [9]: from sklearn.tree import DecisionTreeClassifier
dt=DecisionTreeClassifier(criterion="entropy",max_depth=None)
model=dt.fit(x,y)
y_pred=model.predict(x)
print(y_pred)
```

```
[0 0 1 1 1 0 1 0 1 1 1 1 1 0]
```

```
In [10]: pd.DataFrame({'Actual':y, 'Predicted':y_pred})
```

```
Out[10]:
```

| | Actual | Predicted |
|----|--------|-----------|
| 0 | 0 | 0 |
| 1 | 0 | 0 |
| 2 | 1 | 1 |
| 3 | 1 | 1 |
| 4 | 1 | 1 |
| 5 | 0 | 0 |
| 6 | 1 | 1 |
| 7 | 0 | 0 |
| 8 | 1 | 1 |
| 9 | 1 | 1 |
| 10 | 1 | 1 |
| 11 | 1 | 1 |
| 12 | 1 | 1 |
| 13 | 0 | 0 |

```
In [11]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y,y_pred)
cm
```

```
Out[11]: array([[5, 0],
               [0, 9]], dtype=int64)
```

```
In [12]: TN=cm[0][0]
FP=cm[0][1]
FN=cm[1][0]
TP=cm[1][1]
```

```
In [13]: from sklearn.metrics import accuracy_score
acs=accuracy_score(y,y_pred)
print("acs :",acs )
```

```
acs : 1.0
```

```
In [14]: print("error :",1-ac)
```

```
error : 0.0
```

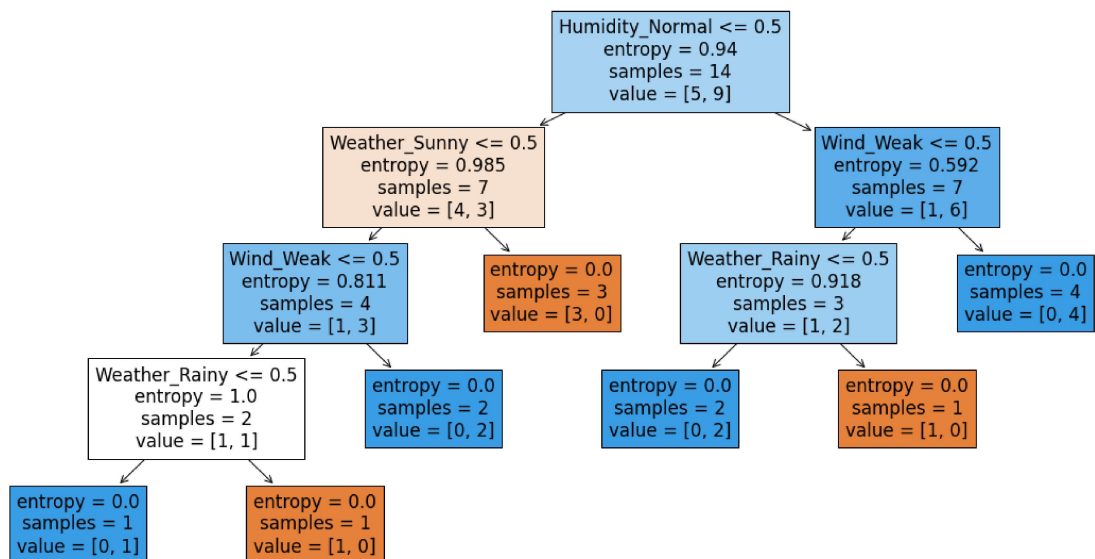
```
In [15]: print("sensitivity :",TP / (TP+FN))
```

```
sensitivity : 1.0
```

```
In [16]: print("specificity :" ,TN / (TN+FP))
```

specificity : 1.0

```
In [22]: import matplotlib.pyplot as plt
from sklearn.tree import plot_tree
plt.figure(figsize=(20,10))
plot_tree(dt,filled=True,feature_names=x.columns)
plt.show()
```



Tree-2

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

Out[1]:

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | DiabetesPedigreeFunctio |
|-----|-------------|---------|---------------|---------------|---------|------|-------------------------|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.62 |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.35 |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.67 |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.16 |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.28 |
| ... | ... | ... | ... | ... | ... | ... | . |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.17 |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.34 |
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.24 |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.34 |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.31 |

768 rows × 9 columns



```
In [4]: y=df['Outcome']
y
```

Out[4]:

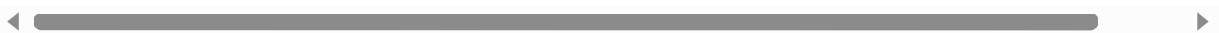
```
0      1
1      0
2      1
3      0
4      1
..
763    0
764    0
765    0
766    1
767    0
Name: Outcome, Length: 768, dtype: int64
```

```
In [5]: x=df.drop("Outcome",axis=1)
x
```

Out[5]:

| | Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | DiabetesPedigreeFunction |
|-----|-------------|---------|---------------|---------------|---------|------|--------------------------|
| 0 | 6 | 148 | 72 | 35 | 0 | 33.6 | 0.62 |
| 1 | 1 | 85 | 66 | 29 | 0 | 26.6 | 0.35 |
| 2 | 8 | 183 | 64 | 0 | 0 | 23.3 | 0.67 |
| 3 | 1 | 89 | 66 | 23 | 94 | 28.1 | 0.16 |
| 4 | 0 | 137 | 40 | 35 | 168 | 43.1 | 2.28 |
| ... | ... | ... | ... | ... | ... | ... | . |
| 763 | 10 | 101 | 76 | 48 | 180 | 32.9 | 0.17 |
| 764 | 2 | 122 | 70 | 27 | 0 | 36.8 | 0.34 |
| 765 | 5 | 121 | 72 | 23 | 112 | 26.2 | 0.24 |
| 766 | 1 | 126 | 60 | 0 | 0 | 30.1 | 0.34 |
| 767 | 1 | 93 | 70 | 31 | 0 | 30.4 | 0.31 |

768 rows × 8 columns



```
In [6]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.8,random_state=
```

```
In [9]: print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(x_test.shape)
```

```
(614, 8)
(154, 8)
(614,)
(154, 8)
```

```
In [12]: from sklearn.tree import DecisionTreeClassifier
dt=DecisionTreeClassifier(criterion='entropy',max_depth=None)
model=dt.fit(x_train,y_train)
y_pred=model.predict(x_test)
print(y_pred)
```

```
[1 0 0 0 0 0 0 0 0 0 1 1 0 1 0 1 0 1 0 0 1 0 1 0 0 0 0 1 0 1 0 1 0 1 1 1 0
 1 0 0 0 0 0 1 0 0 1 0 0 0 0 1 1 0 1 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 1 1 0 0
 0 1 1 1 0 1 1 0 0 1 0 0 1 1 1 0 0 0 1 0 1 1 0 0 1 0 0 0 1 0 0 0 0 1 0 0 1
 0 0 1 0 0 1 0 1 1 0 1 0 0 0 0 0 0 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 1 0 0 0
 0 1 0 1 0 1]
```

```
In [13]: pd.DataFrame({'Actual':y_test,'Predicted':y_pred})
```

```
Out[13]:
```

| | Actual | Predicted |
|-----|--------|-----------|
| 285 | 0 | 1 |
| 101 | 0 | 0 |
| 581 | 0 | 0 |
| 352 | 0 | 0 |
| 726 | 0 | 0 |
| ... | ... | ... |
| 563 | 0 | 1 |
| 318 | 0 | 0 |
| 154 | 1 | 1 |
| 684 | 0 | 0 |
| 643 | 0 | 1 |

154 rows × 2 columns

```
In [14]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
cm
```

```
Out[14]: array([[75, 24],
               [22, 33]], dtype=int64)
```

```
In [15]: TN=cm[0][0]
FP=cm[0][1]
FN=cm[1][0]
TP=cm[1][1]
```

```
In [16]: from sklearn.metrics import accuracy_score
acs=accuracy_score(y_test,y_pred)
print("acs :",acs )
```

acs : 0.7012987012987013

```
In [17]: print("error :",1-ac)
```

error : 0.2987012987012987

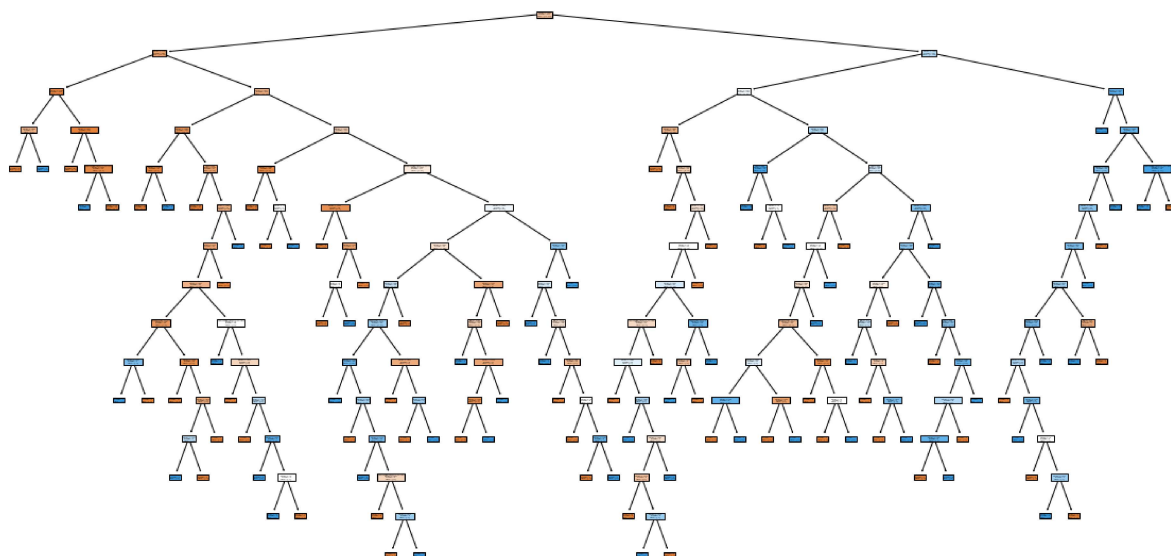
```
In [18]: print("sensitivity :",TP / (TP+FN))
```

sensitivity : 0.6


```
In [19]: print("specificity :" ,TN / (TN+FP))
```

specificity : 0.7575757575757576

```
In [26]: import matplotlib.pyplot as plt
from sklearn.tree import plot_tree
plt.figure(figsize=(20,10))
plot_tree(dt,filled=True,feature_names=x.columns)
plt.show()
```



```
In [ ]:
```