## **Spark Foundation**

## Task 1:Create the Decision Tree classifier and visualize it graphically. $\P$

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```
In [1]: # Importing libraries in python
import pandas as pd
import numpy as np
```

```
In [2]: # Loading Dataset
data=pd.read_csv(r"C:\Users\HP\Downloads\Iris.csv")
data.head()
```

## Out[2]:

	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

```
In [3]: # Data Preprocessing
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
```

	/	0 00-0	
#	Column	Non-Null Count	Dtype
0	Id	150 non-null	int64
1	SepalLengthCm	150 non-null	float64
2	SepalWidthCm	150 non-null	float64
3	PetalLengthCm	150 non-null	float64
4	PetalWidthCm	150 non-null	float64
5	Species	150 non-null	object
dtype	es: float64(4),	int64(1), object	t(1)

memory usage: 7.2+ KB

```
In [4]: data.isnull().sum()
Out[4]: Id
                             0
         SepalLengthCm
                             0
         SepalWidthCm
                             0
         PetalLengthCm
                             0
         PetalWidthCm
                             0
         Species
                             0
         dtype: int64
In [5]: data["Species"].value_counts()
Out[5]: Iris-setosa
                               50
         Iris-versicolor
                               50
         Iris-virginica
                               50
         Name: Species, dtype: int64
In [6]: import seaborn as sns
         sns.heatmap(data.iloc[:,1:4].corr(),annot=True)
Out[6]: <AxesSubplot:>
                                                                                - 1.0
           SepalLengthCm
                                                                                - 0.8
                        1
                                         -0.11
                                                              0.87
                                                                                - 0.6
           SepalWidthCm
                                                                                - 0.4
                      -0.11
                                           1
                                                             -0.42
                                                                                - 0.2
                                                                                - 0.0
           PetalLengthCm
```

```
In [7]: # spliting the dataset
        x=data.iloc[:,:-1].values
        y=data.iloc[:,-1].values
```

-0.42

SepalWidthCm

1

PetalLengthCm

-0.2

-0.4

0.87

SepalLengthCm

```
In [8]: 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_state=
 In [9]: # Fitting model
         from sklearn.tree import DecisionTreeClassifier
         d1=DecisionTreeClassifier()
         d1.fit(x_test,y_test)
 Out[9]: DecisionTreeClassifier()
In [10]: y_pred=d1.predict(x_test)
         y_pred
'Iris-versicolor', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-virginica', 'Iris-virginica',
                'Iris-setosa', 'Iris-versicolor', 'Iris-setosa', 'Iris-versicolor',
                'Iris-versicolor', 'Iris-versicolor', 'Iris-virginica'],
               dtype=object)
In [11]: | from sklearn.metrics import accuracy_score
         acc=accuracy_score(y_pred,y_test)*100
         print(acc)
         100.0
In [12]: # Classification Report
         import sklearn.metrics as metrics
         print(metrics.classification_report(y_test,y_pred))
                         precision
                                      recall f1-score
                                                        support
             Iris-setosa
                              1.00
                                        1.00
                                                 1.00
                                                             10
         Iris-versicolor
                              1.00
                                        1.00
                                                 1.00
                                                             13
          Iris-virginica
                              1.00
                                        1.00
                                                 1.00
                                                              7
                                                 1.00
                                                             30
                accuracy
               macro avg
                              1.00
                                        1.00
                                                 1.00
                                                             30
```

1.00

weighted avg

1.00

1.00

30

```
samples = 30
            value = [10, 13, 7]
           class = Iris-versicolor
                       SepalLengthCm <= 99.5
   gini = 0.0
                             gini = 0.455
 samples = 10
                             samples = 20
value = [10, 0, 0]
                           value = [0, 13, 7]
class = Iris-setosa
                         class = Iris-versicolor
                 gini = 0.0
                                            gini = 0.0
               samples = 13
                                           samples = 7
             value = [0, 13, 0]
                                         value = [0, 0, 7]
                                       class = Iris-virginica
            class = Iris-versicolor
```