## Importing the Libraries

```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import seaborn as sns
   #%matplotlib inline#for encoding
   from sklearn.preprocessing import LabelEncoder#for train test splitting
   from sklearn.model_selection import train_test_split#for decision tree object
   from sklearn.tree import DecisionTreeClassifier#for checking testing results
   from sklearn.metrics import classification_report, confusion_matrix#for visu
   from sklearn.tree import plot_tree
```

Read the files

```
In [2]: from google.colab import files
uploaded = files.upload()
```

Choose Files | no files selected

Upload widget is only available when the cell has

been executed in the current browser session. Please rerun this cell to enable.

Saving Weather.csv to Weather.csv

```
In [3]: df = pd.read_csv("Weather.csv")
   print(df)
```

```
Temp Humidity Windy Play Golf
     Outlook
0
      Rainy
              Hot
                      High False
                                         No
1
      Rainy
              Hot
                      High
                             True
                                         No
2
      Rainy Mild
                      High False
                                         No
3
                    Normal False
      Rainy Cool
                                        Yes
                             True
      Rainy Mild
                    Normal
                                        Yes
5
   Overoact
              Hot
                      High False
                                        Yes
6
   Overoact Cool
                    Normal
                            True
                                        Yes
7
   Overoact Mild
                      High
                            True
                                        Yes
8
             Hot Normal False
   Overoact
                                        Yes
9
      Sunny Mild
                      High False
                                        Yes
      Sunny Cool
10
                    Normal False
                                        Yes
11
       Sunny Cool
                    Normal
                             True
                                         No
12
       Sunny Mild
                    Normal False
                                        Yes
13
      Sunny Mild
                      High
                             True
                                         No
```

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

4/3/23, 8:14 PM Decision\_tree\_weather

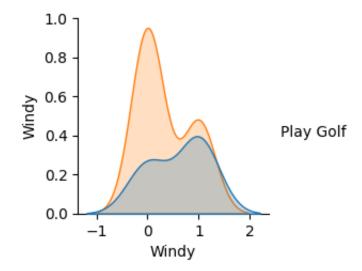
> <class 'pandas.core.frame.DataFrame'> RangeIndex: 14 entries, 0 to 13 Data columns (total 5 columns): # Column Non-Null Count Dtype 0 Outlook 14 non-null object 1 Temp 14 non-null object 2 Humidity 14 non-null object 3 Windy 14 non-null bool object Play Golf 14 non-null dtypes: bool(1), object(4) memory usage: 590.0+ bytes

Check if there are missing values

```
In [5]:
        df.isnull().any()
        Outlook
                      False
Out[5]:
        Temp
                      False
        Humidity
                      False
        Windy
                      False
        Play Golf
                      False
        dtype: bool
In [6]:
        # let's plot pair plot to visualise the attributes all at once
```

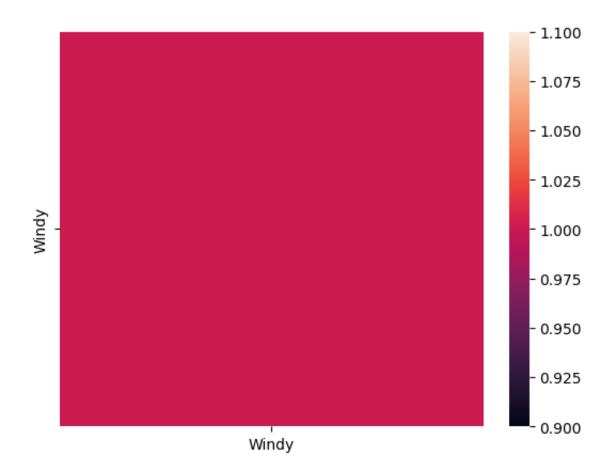
sns.pairplot(data=df, hue = 'Play Golf')

<seaborn.axisgrid.PairGrid at 0x7fe9ad8ba850> Out[6]:



```
In [7]:
        # correlation matrix
        sns.heatmap(df.corr())
```

<Axes: > Out[7]:



## **Data Preprocessing**

Identify the target columns

```
In [8]: #separate the target variable(y) and features(X) as follows
  target = df['Play Golf']
  df1 = df.copy()
  df1 = df1.drop('Play Golf', axis =1)
```

Convert the variables to numeric values

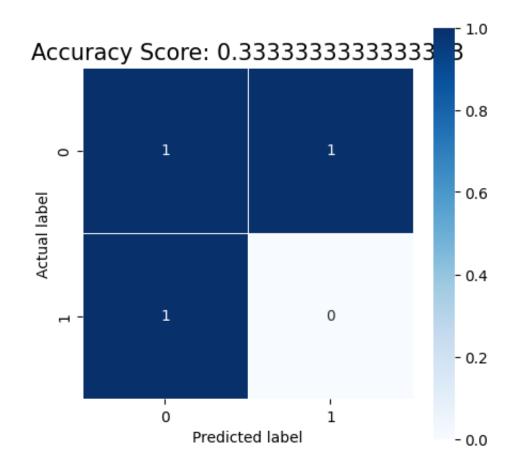
```
In [9]: # Data has categorical variables stored in it we will encode it in numeric v
X = dfl.apply(LabelEncoder().fit_transform)
print(X)
```

```
Humidity
     Outlook
                  Temp
                                        Windy
0
              1
                                    0
                                              0
                      1
1
              1
                      1
                                    0
                                              1
                                    0
2
              1
                      2
                                              0
3
              1
                      0
                                    1
                                              0
4
              1
                      2
                                    1
                                              1
5
              0
                      1
                                    0
                                              0
6
              0
                      0
                                    1
                                              1
7
              0
                      2
                                    0
                                              1
                                              0
8
              0
                      1
                                    1
9
              2
                      2
                                    0
                                              0
10
              2
                      0
                                    1
                                              0
              2
                      0
11
                                    1
                                              1
              2
                      2
                                              0
12
                                    1
13
              2
                      2
                                    0
                                              1
```

```
In [10]:
         target
                No
Out[10]:
         1
                No
         2
                No
         3
               Yes
         4
               Yes
         5
               Yes
         6
               Yes
         7
               Yes
         8
               Yes
         9
               Yes
         10
               Yes
         11
                No
         12
               Yes
         13
                No
         Name: Play Golf, dtype: object
In [11]: #label encoding
         from sklearn import preprocessing
         le = preprocessing.LabelEncoder()
         target = le.fit_transform(target)
         target
         array([0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0])
Out[11]:
In [12]: y = target
         print(y)
         In [13]: from sklearn.model_selection import KFold, cross_val_score, train_test_split
         # Splitting the data - 80:20 ratio
         X_train, X_test, y_train, y_test = train_test_split(X , y, test_size = 0.2,
         print("Testing split input- ", X_test.shape)
         Testing split input- (3, 4)
```

## **Modeling Tree and testing it**

```
In [14]: from sklearn import tree
         # Defining the decision tree algorithmdtree=DecisionTreeClassifier()
         dtree = tree.DecisionTreeClassifier()
         dtree.fit(X_train,y_train)
         print('Decision Tree Classifier Created')
         Decision Tree Classifier Created
In [15]:
        # Predicting the values of test data
         y pred = dtree.predict(X test)
         print("Classification report - \n", classification_report(y_test,y_pred))
         Classification report -
                         precision
                                      recall f1-score
                                                         support
                                                 0.50
                     0
                             0.50
                                       0.50
                                                               2
                     1
                             0.00
                                       0.00
                                                 0.00
                                                               1
                                                 0.33
                                                               3
             accuracy
                             0.25
                                       0.25
                                                 0.25
                                                               3
            macro avq
         weighted avg
                             0.33
                                       0.33
                                                 0.33
                                                               3
In [16]:
         cm = confusion_matrix(y_test, y_pred)
         plt.figure(figsize=(5,5))
         sns.heatmap(data=cm,linewidths=.5, annot=True, square = True, cmap = 'Blues'
         plt.ylabel('Actual label')
         plt.xlabel('Predicted label')
         all sample_title = 'Accuracy Score: {0}'.format(dtree.score(X_test, y_test))
         plt.title(all_sample_title, size = 15)
         Text(0.5, 1.0, 'Accuracy Score: 0.3333333333333333')
Out[16]:
```



## Visualizing the decision tree

In [17]: # Visualising the graph without the use of graphvizplt.figure(figsize = (20,
dec\_tree = plot\_tree(decision\_tree=dtree, feature\_names = df1.columns, class

