

### What is our GOAL for this MODULE?

The goal of this module is to explore the concept of Decision Tree.

## What did we ACHIEVE in the class TODAY?

We explored the concept of Decision Tree , wrote our own Decision Tree algorithm and created the charts for same..

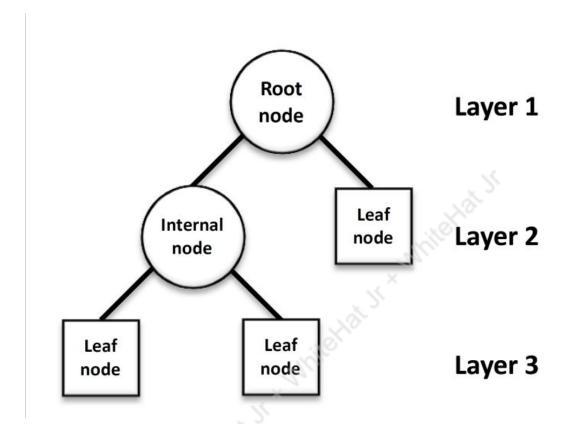
## Which CONCEPTS/CODING BLOCKS did we cover today?

- Decision Tree algorithm
- Usage of **pydotplus** module to turn the data into chart images.



### How did we DO the activities?

- 1. We explored the concept of Decision Tree.
  - We saw the different nodes of the Decision Tree.



- 2. We used the data of diabetes patients for our algorithm.
- 3. We uploaded the data and then created the data frames of it.

```
#Uploading the csv
from google.colab import files
data_to_load = files.upload()
```



```
import pandas as pd
#Column Name
col names = ['pregnant', 'glucose', 'bp', 'skin', 'insulin', 'bmi', 'pedigree', 'age', 'label']
df = pd.read csv("diabetes.csv", names=col names).iloc[1:]
print(df.head())
  pregnant glucose
                   bp skin insulin
                                     bmi pedigree age label
              148
                   72
                        35
                                   33.6
                                           0.627
                                                  50
                                0 26.6
2
                                           0.351 31
               85
                   66
                        29
                                                         0
         1
3
              183 64
                        0
                                0 23.3
                                           0.672 32
4
               89 66
                        23
                                94 28.1
                                           0.167 21
                                                         0
         1
5
         0
               137
                   40
                        35
                               168
                                   43.1
                                           2.288
features = ['pregnant', 'insulin', 'bmi', 'age', 'glucose', 'bp', 'pedigree']
X = df[features]
y = df.label
```

4. Then we split the data to train, test and then fit it into the model.

```
[ ] from sklearn.tree import DecisionTreeClassifier
    from sklearn.model_selection import train_test_split
    from sklearn import metrics

#splitting data in training and testing
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=1)

#Initialising the Decision Tree Model
    clf = DecisionTreeClassifier()

#Fitting the data into the model
    clf = clf.fit(X_train,y_train)

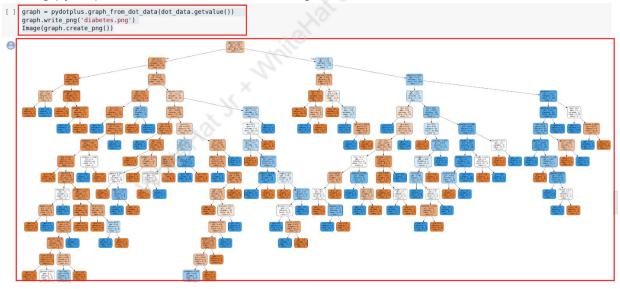
#Calculating the accuracy of the model
    y_pred = clf.predict(X_test)
    print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
Accuracy: 0.666666666666666666
```

**5.** Using the graph viz function we created the representation of the data.



```
from sklearn.tree import export_graphviz
from sklearn.externals.six import String
          from IPython.display import Image
          import pydotplus
          dot data = StringIO() #Where we will store the data from our decision tree classifier as text.
          export graphviz(clf, out file=dot data, filled=True, rounded=True, special characters=True, feature names=features, class names=['0','1'])
          print(dot data.getvalue())
       digraph Tree {
   node [shape=box, style="filled, rounded", color="black", fontname=helvetica];
   edge [fontname=helvetica];
   0 [label=<glucose &le; 129.5<br/>cr/>gini = 0.449<br/>cbr/>samples = 537<br/>cbr/>value = [354, 183]<br/>cbr/>class = 0>, fillcolor="#f2c29f"];
   1 [label=<br/>cbi de; 26.3<br/>cbr/>qini = 0.329<br/>cbr/>samples = 357<br/>cbr/>value = [283, 74]<br/>cbr/>class = 0>, fillcolor="#ec26d"];
   0 -> 1 [labeldistance=2.5, labelangle=45, headlabel="True"];
   2 [label=<br/>cbmi &le; 9.1<br/>cbr/>gini = 0.06<br/>cbr/>samples = 97<br/>cbr/>value = [94, 3]<br/>cbr/>class = 0>, fillcolor="#e6853f"];
   1 -> 2 [value = value = valu
0
             [label=<age &le; 28.0<br/>jini = 0.444<br/>samples = 6<br/>value = [4, 2]<br/>class = 0>, fillcolor="#f2c09c"];
        4 [label=<gini = 0.0<br/>br/>samples = 4<br/>br/>value = [4, 0]<br/>class = 0>, fillcolor="#e58139"] ;
             [label=<gini = 0.0<br/>samples = 2<br/>value = [0, 2]<br/>class = 1>, fillcolor="#399de5"];
        5 [Jabel=<pedigree &le; 0.669<br/>spini = 0.022<br/>samples = 91<br/>br/>value = [90, 1]<br/>br/>class = 0>, fillcolor="#e5823b"];
             [label=<gini=0.0 < br/> samples=76 < br/> value=[76, 0] < br/> class=0>, fillcolor="#e58139"] ;
            [label=spedigree ≤ 0.705<br/>gini = 0.124<br/>samples = 15<br/>value = [14, 1]<br/>class = 0>, fillcolor="#e78a47"] ;
        9 [label=<qini = 0.0<br/>samples = 1<br/>value = [0, 1]<br/>class = 1>, fillcolor="#399de5"];
        10 [label=<gini = 0.0<br/>br/>samples = 14<br/>br/>value = [14, 0]<br/>cbr/>class = 0>, fillcolor="#e58139"] ; 8 \rightarrow 10 ;
         11 [label=<age &le; 27.5<br/>br/>gini = 0.397<br/>br/>samples = 260<br/>br/>value = [189, 71]<br/>class = 0>, fillcolor="#efb083"];
         12 [label=<bmi &le: 45.4<br/>pr/>qini = 0.243<br/>samples = 120<br/>br/>value = [103, 17]<br/>class = 0>, fillcolor="#e9965a"];
             :-> 12 ,
| Clabel=<br/>clap &le; 12.0<br/>gini = 0.212<br/>samples = 116<br/>value = [102, 14]<br/>class = 0>, fillcolor="#e99254"] ; -> 13 ;
```

**6.** Using pydotplus we created the chart image of the data.



7. We trimmed the data and then again created the image chart of that data.

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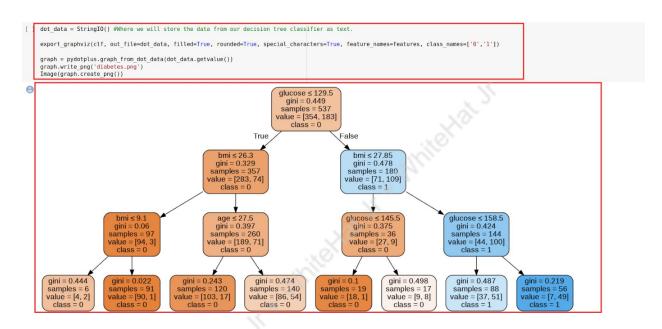


```
[ ] clf = DecisionTreeClassifier(max_depth=3)

clf = clf.fit(X_train,y_train)

y_pred = clf.predict(X_test)
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

Accuracy: 0.75757575757576



WE concluded that almost 75% accuracy that a person who's

**Glucose** is greater than 129.5 and, **BMI** is greater than 27.85 Is more prone to be a Diabetes Patient.

#### What's NEXT?

In the next class, we will explore more of machine learning.

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# PRO-C118



## **EXTEND YOUR KNOWLEDGE:**

Learn more about the decision tree from the following link: <a href="https://scikit-learn.org/stable/modules/tree.html">https://scikit-learn.org/stable/modules/tree.html</a>