## Untitled108

July 2, 2023

```
[1]: ## import the required library
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

## 1 Module 8: Decision Tree

Assignment Contact us: support@intellipaat.com / © Copyright Intellipaat / All rights reserved Intel iPaat Python for Data Science Certification Course Problem Statement: You work in XYZ Company as a Python Data Scientist. The company officials have collected some data on salaries based on year of experience and wish for you to create a model from it. Dataset: diabetes.csv Tasks To Be Performed: 1. Load the dataset using pandas 2. Extract data from outcome column is a variable named Y 3. Extract data from every column except outcome column in a variable named X 4. Divide the dataset into two parts for training and testing in 70% and 30% proportion 5. Create and train Decision Tree Model on training set 6. Make predictions based on the testing set using the trained model 7. Check the performance by calculating the confusion matrix and accuracy score of the model

```
[2]: ##Load the dataset using pandas
data=pd.read_csv(r'C:/Users/Vikas/Desktop/diabetes-1.csv')
```

[3]: data

[3]:	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	
	•••	•••	•••		•••		
763	10	101	76	48	180	32.9	
764	2	122	70	27	0	36.8	
765	5	121	72	23	112	26.2	
766	1	126	60	0	0	30.1	
767	1	93	70	31	0	30.4	

		DiabetesPedi	greeFunction	n Age	Outcor	m 🛆				
	0	Diabetesi edi	0.62	•	outcon	1				
	1		0.35			0				
	2		0.67			1				
	3		0.16			0				
	4		2.28			1				
	•					1				
	763		0.17		•••	0				
	764		0.34			0				
	765		0.24			0				
	766		0.349			1				
	767		0.31			0				
			0.00							
	[768	rows x 9 col	umns]							
[31]:	data	head(5)								
[04]			llugger Di	D	(7)	-i-Thisle	Tm === 3.4.	DMT	,	
[31]:		_		odPress		kinThickness	Insulin		\	
	0	6	148		72	35	0	33.6		
	1	1	85		66	29	0	26.6		
	2	8	183		64	0	0	23.3		
	3	1	89		66	23	94	28.1		
	4	0	137		40	35	168	43.1		
	ח	iabetesPedigr	eeFunction	Age C	Outcome					
	0	14000001 04161	0.627	50	1					
	1		0.351	31	0					
	2		0.672	32	1					
	3		0.167	21	0					
	4		2.288	33	1					
[32]:	data	tail(5)								
[32]:	<b>7.65</b>	-		loodPre		SkinThicknes				
	763	10	101		76		l8 18			
	764	2	122		70			0 36.8		
	765	5	121		72	2	23 11			
	766	1	126		60			0 30.1		
	767	1	93		70	3	31	0 30.4	:	
		Dishar - D. 11		^	O+ -					
	762	DiabetesPedi	•	_	Outcom					
	763		0.17			0				
	764		0.34	0 27		0				

0.245

0.349

0.315

```
[33]: data.shape
[33]: (768, 9)
[28]:
     data.columns
[28]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
             'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
            dtype='object')
[29]: data.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
          Glucose
                                     768 non-null
                                                      int64
      1
          BloodPressure
                                     768 non-null
                                                      int64
          SkinThickness
                                     768 non-null
                                                      int64
      4
          Insulin
                                     768 non-null
                                                      int64
      5
          BMI
                                     768 non-null
                                                      float64
      6
          DiabetesPedigreeFunction 768 non-null
                                                      float64
      7
                                     768 non-null
                                                      int64
          Age
          Outcome
                                     768 non-null
                                                      int64
     dtypes: float64(2), int64(7)
     memory usage: 54.1 KB
[30]: data.isnull().sum()
[30]: Pregnancies
                                   0
      Glucose
                                   0
      BloodPressure
                                   0
      SkinThickness
                                   0
      Insulin
                                   0
      BMI
                                   0
      DiabetesPedigreeFunction
                                   0
                                   0
      Age
      Outcome
                                   0
      dtype: int64
[12]: ##Extract data from outcome column is a variable named Y
      y=pd.DataFrame(data.iloc[:,-1])
 []:
[13]: y
```

[13]:		Outcome
	0	1
	1	0
	2	1
	3	0
	4	1
		•••
	763	0
	764	0
	765	0
	766	1
	767	0

[768 rows x 1 columns]

[15]: ##Extract data from every column except outcome column in a variable named X x=pd.DataFrame(data.iloc[:,:-1])

[16]: x

[16]:	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	
	•••	•••	•••				
763	10	101	76	48	180	32.9	
764	2	122	70	27	0	36.8	
765	5	121	72	23	112	26.2	
766	1	126	60	0	0	30.1	
767	1	93	70	31	0	30.4	

	DiabetesPedigreeFunction			
0	0.627	50		
1	0.351	31		
2	0.672	32		
3	0.167	21		
4	2.288	33		
763	0.171	63		
764	0.340	27		
765	0.245	30		
766	0.349	47		
767	0.315	23		

[768 rows x 8 columns]

```
[19]: | ##Divide the dataset into two parts for training and testing in 70% and 30% |
       \hookrightarrowproportion
      from sklearn.model selection import train test split
      x_train, x_test, y_train, y_test = train_test_split(
      x, y, test_size=0.30, random_state=0)
[20]: from sklearn.tree import DecisionTreeClassifier
      clf=DecisionTreeClassifier()
      clf=clf.fit(x_train,y_train)
[22]: clf.get_params()
[22]: {'ccp_alpha': 0.0,
       'class weight': None,
       'criterion': 'gini',
       'max depth': None,
       'max_features': None,
       'max_leaf_nodes': None,
       'min_impurity_decrease': 0.0,
       'min_samples_leaf': 1,
       'min_samples_split': 2,
       'min_weight_fraction_leaf': 0.0,
       'random_state': None,
       'splitter': 'best'}
[25]: ##Make predictions based on the testing set using the trained model
      y_pred=clf.predict(x_test)
      y_pred
[25]: array([0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0,
             0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1,
             0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1,
             0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
             1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0,
             0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0,
             0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
             1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0,
             1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0,
             0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0,
             0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 0], dtype=int64)
[26]: ## confusion matrixc
      from sklearn.metrics import confusion_matrix
      cm=confusion_matrix(y_test,y_pred)
      print('confusionmatrix:\n',cm)
     confusionmatrix:
      [[123 34]
```

## [ 26 48]]

```
[27]: ##accuracy score
from sklearn.metrics import accuracy_score
print("Accuracy:",accuracy_score(y_test,y_pred))
Accuracy: 0.7402597402597403
```