

Importing Required Libraries

```
In [49]: import pandas as pd
import numpy as np
import seaborn as sns
from matplotlib import pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
import warnings# used for removing warnings
warnings.filterwarnings('ignore')
from sklearn.tree import plot_tree
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
```

Impoting the data

```
In [3]: affair = pd.read_csv("affairs.csv")
affair
```

```
Out[3]:
```

	Unnamed: 0	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel
0	1	0	0	0	0	0	1	0	0	0	
1	2	0	0	0	0	0	1	0	0	0	
2	3	3	0	0	0	0	1	0	0	0	
3	4	0	1	0	0	0	1	0	1	0	
4	5	3	1	0	0	0	0	1	0	0	
...	
596	597	0	0	0	0	0	1	0	0	0	
597	598	1	1	0	0	0	0	1	0	0	
598	599	0	1	0	1	0	0	0	0	1	
599	600	0	1	0	0	0	1	0	0	0	
600	601	0	1	0	0	0	1	0	0	1	

601 rows × 19 columns



Initial Analysis

```
In [4]: affair.head()
```

```
Out[4]:
```

	Unnamed: 0	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel
0	1	0	0	0	0	0	1	0	0	0	1
1	2	0	0	0	0	0	1	0	0	0	0

	Unnamed: 0	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel
2	3	3	0	0	0	0	1	0	0	0	1
3	4	0	1	0	0	0	1	0	1	0	0
4	5	3	1	0	0	0	0	1	0	0	1

In [5]: `affair.shape`

Out[5]: (601, 19)

In [6]: `affair.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 601 entries, 0 to 600
Data columns (total 19 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Unnamed: 0      601 non-null    int64
1   naffairs        601 non-null    int64
2   kids            601 non-null    int64
3   vryunhap        601 non-null    int64
4   unhap           601 non-null    int64
5   avgmarr         601 non-null    int64
6   hapavg          601 non-null    int64
7   vryhap          601 non-null    int64
8   antirel         601 non-null    int64
9   notrel          601 non-null    int64
10  slghtrel        601 non-null    int64
11  smerel          601 non-null    int64
12  vryrel          601 non-null    int64
13  yrsmarr1        601 non-null    int64
14  yrsmarr2        601 non-null    int64
15  yrsmarr3        601 non-null    int64
16  yrsmarr4        601 non-null    int64
17  yrsmarr5        601 non-null    int64
18  yrsmarr6        601 non-null    int64
dtypes: int64(19)
memory usage: 89.3 KB
```

In [7]: `affair.isna().sum()`

```
Out[7]: Unnamed: 0      0
naffairs      0
kids          0
vryunhap      0
unhap         0
avgmarr       0
hapavg        0
vryhap        0
antirel       0
notrel        0
slghtrel      0
smerel        0
vryrel        0
yrsmarr1      0
yrsmarr2      0
yrsmarr3      0
yrsmarr4      0
yrsmarr5      0
```

yrs marr6 0
dtype: int64

```
In [8]: affair.describe()
```

Out[8]:

	Unnamed: 0	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap
count	601.000000	601.000000	601.000000	601.000000	601.000000	601.000000	601.000000	601.000000
mean	301.000000	1.455907	0.715474	0.026622	0.109817	0.154742	0.322795	0.386000
std	173.638033	3.298758	0.451564	0.161111	0.312922	0.361960	0.467935	0.487200
min	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	151.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
50%	301.000000	0.000000	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000
75%	451.000000	0.000000	1.000000	0.000000	0.000000	0.000000	1.000000	1.000000
max	601.000000	12.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000

```
In [10]: affair.dtypes
```

Out[10]: Unnamed: 0 int64
naffairs int64
kids int64
vryunhap int64
unhap int64
avgmarr int64
hapavg int64
vryhap int64
antirel int64
notrel int64
slghtrel int64
smere1 int64
vryrel int64
yrs marr1 int64
yrs marr2 int64
yrs marr3 int64
yrs marr4 int64
yrs marr5 int64
yrs marr6 int64
dtype: object

```
In [11]: affair.dropna(axis = 0,inplace = True)
```

```
In [12]: affair.head(10)
```

Out[12]:

	Unnamed: 0	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel
0	1	0	0	0	0	0	1	0	0	0	1
1	2	0	0	0	0	0	1	0	0	0	0
2	3	3	0	0	0	0	1	0	0	0	1
3	4	0	1	0	0	0	1	0	1	0	0

	Unnamed: 0	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel
4	5	3	1	0	0	0	0	1	0	0	1
5	6	0	1	0	0	0	0	1	0	0	0
6	7	0	0	0	0	1	0	0	0	1	0
7	8	0	0	0	0	0	0	1	0	1	0
8	9	7	1	0	1	0	0	0	0	0	0
9	10	0	0	0	0	1	0	0	0	1	0

In [13]: `affair.shape`

Out[13]: (601, 19)

In [14]: `del affair['Unnamed: 0']`

In [15]: `affair.head()`

	naffairs	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel	smerel	vry
0	0	0	0	0	0	1	0	0	0	1	0	
1	0	0	0	0	0	1	0	0	0	0	0	1
2	3	0	0	0	0	1	0	0	0	0	1	0
3	0	1	0	0	0	1	0	1	0	0	0	0
4	3	1	0	0	0	0	1	0	0	0	1	0

Model Building

Seperate input features

In [16]: `x = affair.drop(labels = 'naffairs' , axis = 1)`
`y = affair[['naffairs']]`

In [19]: `x`

	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel	smerel	vryrel	yr
0	0	0	0	0	1	0	0	0	1	0	0	
1	0	0	0	0	1	0	0	0	0	1	0	
2	0	0	0	0	1	0	0	0	1	0	0	
3	1	0	0	0	1	0	1	0	0	0	0	

	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel	smerel	vryrel	yr
4	1	0	0	0	0	1	0	0	1	0	0	
...
596	0	0	0	0	1	0	0	0	0	1	0	
597	1	0	0	0	0	1	0	0	1	0	0	
598	1	0	1	0	0	0	0	1	0	0	0	
599	1	0	0	0	1	0	0	0	1	0	0	
600	1	0	0	0	1	0	0	1	0	0	0	

601 rows × 17 columns



In [20]:

```
y
```

Out[20]:

	naffairs
0	0
1	0
2	3
3	0
4	3
...	...
596	0
597	1
598	0
599	0
600	0

601 rows × 1 columns

In [22]:

```
x_train,x_test,y_train,y_test = train_test_split(x,y,random_state=12,test_size=0.2)
```

In [23]:

```
x_train
```

Out[23]:

	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel	smerel	vryrel	yr
293	1	0	0	1	0	0	0	0	1	0	0	
377	1	0	0	0	1	0	0	0	0	1	0	
541	0	0	0	0	0	1	1	0	0	0	0	
543	1	0	0	0	1	0	0	0	0	1	0	
5	1	0	0	0	0	1	0	0	0	0	1	
...

	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel	smerel	vryrel	yr
432	1	0	0	1	0	0	0	0	0	1	0	
259	1	0	0	1	0	0	1	0	0	0	0	
241	0	0	0	0	1	0	0	0	1	0	0	
253	1	0	0	0	0	1	0	1	0	0	0	
390	1	0	0	0	1	0	0	0	1	0	0	

480 rows × 17 columns



In [24]:

x_test

Out[24]:

	kids	vryunhap	unhap	avgmarr	hapavg	vryhap	antirel	notrel	slghtrel	smerel	vryrel	yr
14	1	0	1	0	0	0	0	0	1	0	0	
262	1	0	0	1	0	0	0	1	0	0	0	
453	1	0	0	1	0	0	0	0	1	0	0	
326	1	0	0	1	0	0	1	0	0	0	0	
210	1	0	0	1	0	0	0	1	0	0	0	
...
558	1	0	0	0	0	1	0	0	1	0	0	
307	1	0	0	0	1	0	0	0	0	1	0	
368	1	0	1	0	0	0	0	1	0	0	0	
527	1	0	0	0	0	1	0	0	0	1	0	
553	1	0	0	1	0	0	0	1	0	0	0	

121 rows × 17 columns



In [26]:

x_train.shape,y_train.shape

Out[26]: ((480, 17), (480, 1))

In [27]:

x_test.shape , y_test.shape

Out[27]: ((121, 17), (121, 1))

Model Training

In [34]:

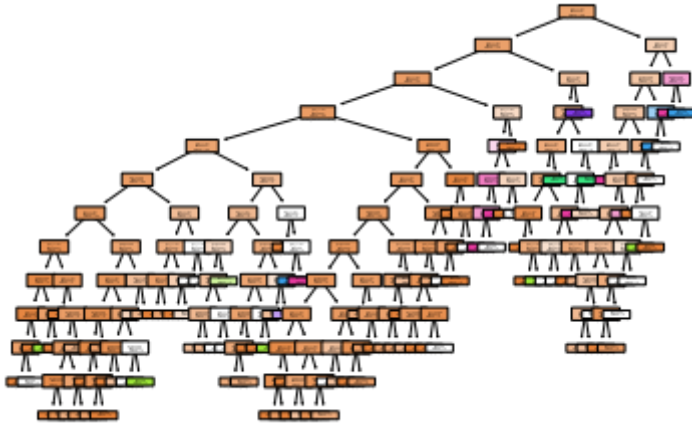
```
logic_model = LogisticRegression()
logic_model.fit(x_train,y_train)

dt_model = DecisionTreeClassifier()
dt_model.fit(x_train,y_train)
```

```
Out[34]: DecisionTreeClassifier()
```

```
In [35]: ## preparing a plot figure

plot_tree(dt_model,rounded = True,filled = True)
plt.figure(figsize=(16,10))
plt.show()
```



<Figure size 1152x720 with 0 Axes>

Model Testing

training data

```
In [37]: # y_pred_train = logic_model.predict(x_train)
# y_pred_train
```

```
In [38]: y_pred_train = dt_model.predict(x_train)
          y_pred_train
```

```
Out[38]: array([[ 1,  0, 12,  0,  0,  0,  0,  0,  0,  0, 12,  0,  0,  0,  0,  0,  7,
  0,  0,  0,  0,  0,  2,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  7,  0,  0,  0,  0,  0,  0,  7,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0, 12,  0,  0,  0,  0,  0,  0,  0,  0,  0, 12,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  1,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  2,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  7,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  1,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0, 12,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  3, 12,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  2,  0,  0,  0,
  0,  0,  0,  0,  0, 12,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  1,  0,  3,  0,  0,  7,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0, 12,  3,  0,  0,  0,  0,  0,  0,  0,
  0,  7,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0])
```

```
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 12, 1, 0,
7, 0, 0, 0], dtype=int64)
```

```
In [40]: accuracy_score(y_train,y_pred_train)
```

```
Out[40]: 0.7979166666666667
```

```
In [41]: confusion_matrix(y_train,y_pred_train)
```

```
Out[41]: array([[355,  0,  0,  0,  1,  1],
 [ 22,  7,  0,  0,  0,  0],
 [ 11,  1,  3,  0,  0,  0],
 [ 13,  0,  0,  3,  0,  0],
 [ 26,  0,  0,  0,  6,  0],
 [ 22,  0,  0,  0,  0,  9]], dtype=int64)
```

```
In [42]: print(classification_report(y_train,y_pred_train))
```

	precision	recall	f1-score	support
0	0.79	0.99	0.88	357
1	0.88	0.24	0.38	29
2	1.00	0.20	0.33	15
3	1.00	0.19	0.32	16
7	0.86	0.19	0.31	32
12	0.90	0.29	0.44	31
accuracy			0.80	480
macro avg	0.90	0.35	0.44	480
weighted avg	0.82	0.80	0.75	480

```
In [44]: # y_pred_test = logic_model.predict(x_test)
# y_pred_test
```

```
y_pred_test = dt_model.predict(x_test)
y_pred_test
```

```
Out[44]: array([12,  0,  1,  7,  1,  0,  0, 12,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  1,  0,  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  2,  0,  0,  0,
  0,  0,  0,  7,  0,  0,  0,  0,  0,  0,  0,  0,  1,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0, 12,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  3,
  0,  1], dtype=int64)
```

```
In [45]: accuracy_score(y_test,y_pred_test)
```

```
Out[45]: 0.71900826446281
```

```
In [46]: confusion_matrix(y_test,y_pred_test)
```

```
Out[46]: array([[85,  6,  1,  0,  1,  1],
 [ 5,  0,  0,  0,  0,  0],
 [ 2,  0,  0,  0,  0,  0],
 [ 3,  0,  0,  0,  0,  0],
```



```
[ 9,  0,  0,  1,  0,  0],
 [ 4,  0,  0,  0,  1,  2]], dtype=int64)
```

```
In [48]: print(classification_report(y_test,y_pred_test))
```

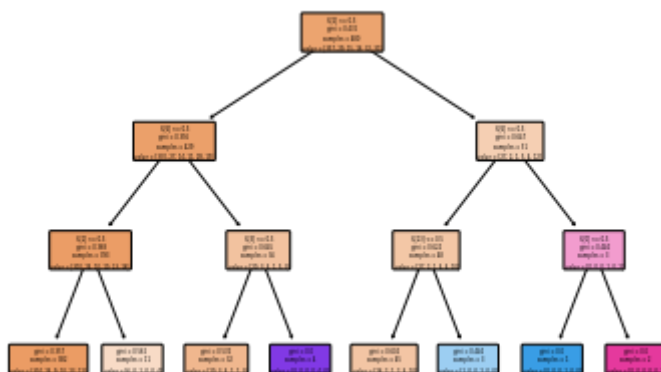
	precision	recall	f1-score	support
0	0.79	0.90	0.84	94
1	0.00	0.00	0.00	5
2	0.00	0.00	0.00	2
3	0.00	0.00	0.00	3
7	0.00	0.00	0.00	10
12	0.67	0.29	0.40	7
accuracy			0.72	121
macro avg	0.24	0.20	0.21	121
weighted avg	0.65	0.72	0.68	121

Model Improvement with decision Tree

```
In [51]: dt_model = DecisionTreeClassifier( criterion='gini',max_depth=3)
dt_model.fit(x_train,y_train)
```

```
Out[51]: DecisionTreeClassifier(max_depth=3)
```

```
In [62]: plot_tree(dt_model,filled = True,rounded = True)
plt.figure(figsize = (20,12))
plt.show()
```



<Figure size 1440x864 with 0 Axes>

```
In [ ]:
```