Decision Trees Algorithm

In [1]:
```
#Exp no.:11
```

In [2]:
```
#Aim : Understanding Decision Trees Algorithm
```

In [3]:
```
#Name:Vedant M.Padole
#Roll no:42
#Sec:C
#Subject:ET1
#Date:
```

Importing The Libraries

In [4]:
```
import pandas as pd
import numpy as np
```

In [7]:
```
import os
```

In [8]:
```
os.getcwd()
```

Out[8]:
```
'C:\\Users\\DELL'
```

In [10]:
```
os.chdir('C:\\Users\\DELL\\Desktop')
```

In [11]:
```
data=pd.read_csv("heart.csv")
```

In [12]:
```
data.head()
```

Out[12]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	target
0	52	1	0	125	212	0	1	168	0	1.0	2	2	3	0
1	53	1	0	140	203	1	0	155	1	3.1	0	0	3	0
2	70	1	0	145	174	0	1	125	1	2.6	0	0	3	0
3	61	1	0	148	203	0	1	161	0	0.0	2	1	3	0
4	62	0	0	138	294	1	1	106	0	1.9	1	3	2	0

In [13]:
```
data.tail()
```

Out[13]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	target
1020	59	1	1	140	221	0	1	164	1	0.0	2	0	2	1
1021	60	1	0	125	258	0	0	141	1	2.8	1	1	3	0
1022	47	1	0	110	275	0	0	118	1	1.0	1	1	2	0
1023	50	0	0	110	254	0	0	159	0	0.0	2	0	2	1
1024	54	1	0	120	188	0	1	113	0	1.4	1	1	3	0

In [14]:
```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1025 entries, 0 to 1024
```

```
Data columns (total 14 columns):
 #   Column    Non-Null Count Dtype
--- ------    -------------- -----
0   age       1025 non-null  int64
1   sex       1025 non-null  int64
2   cp        1025 non-null  int64
3   trestbps  1025 non-null  int64
4   chol      1025 non-null  int64
5   fbs       1025 non-null  int64
6   restecg   1025 non-null  int64
7   thalach   1025 non-null  int64
8   exang     1025 non-null  int64
9   oldpeak   1025 non-null  float64
10  slope     1025 non-null  int64
11  ca        1025 non-null  int64
12  thal      1025 non-null  int64
13  target    1025 non-null  int64
dtypes: float64(1), int64(13)
memory usage: 112.2 KB
```

In [15]:
```
data.describe()
```

Out[15]:

	age	sex	cp	trestbps	chol	fbs	restecg	
count	1025.000000	1025.000000	1025.000000	1025.000000	1025.00000	1025.000000	1025.000000	102
mean	54.434146	0.695610	0.942439	131.611707	246.00000	0.149268	0.529756	14
std	9.072290	0.460373	1.029641	17.516718	51.59251	0.356527	0.527878	2
min	29.000000	0.000000	0.000000	94.000000	126.00000	0.000000	0.000000	7
25%	48.000000	0.000000	0.000000	120.000000	211.00000	0.000000	0.000000	13
50%	56.000000	1.000000	1.000000	130.000000	240.00000	0.000000	1.000000	15
75%	61.000000	1.000000	2.000000	140.000000	275.00000	0.000000	1.000000	16
max	77.000000	1.000000	3.000000	200.000000	564.00000	1.000000	2.000000	20

In [16]:
```
data.shape
```

Out[16]: (1025, 14)

In [18]:
```
data.size
```

Out[18]: 14350

In [19]:
```
data.ndim
```

Out[19]: 2

Data preprocessing _ data cleaning _missing value treatment

In [20]:
```
# check Missing Value by record
data.isna()
```

Out[20]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal
0	False	False	False	False	False	False	False	False	False	False	False	False	False

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal
1	False	False	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False	False	False
...	...	...	...	...	...	...	...	...	...	...	...	...	...
1020	False	False	False	False	False	False	False	False	False	False	False	False	False
1021	False	False	False	False	False	False	False	False	False	False	False	False	False
1022	False	False	False	False	False	False	False	False	False	False	False	False	False
1023	False	False	False	False	False	False	False	False	False	False	False	False	False
1024	False	False	False	False	False	False	False	False	False	False	False	False	False

1025 rows × 14 columns

In [21]:
```
data.isna().any()
```

Out[21]:
```
age         False
sex         False
cp          False
trestbps    False
chol        False
fbs         False
restecg     False
thalach     False
exang       False
oldpeak     False
slope       False
ca          False
thal        False
target      False
dtype: bool
```

Independent and Dependent Variables

In [22]:
```
x=data.drop("target", axis=1)
y=data["target"]
```

Splitting of DataSet into train and Test

In [23]:
```
# Splitting the data into training and testing data sets
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_stat
```

Decision Trees Algorithm

In [24]:
```
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score
```

In [25]:
```
dt=DecisionTreeClassifier()
```

In [26]:
```
dt.fit(x_train, y_train)
```

Out[26]:
```
DecisionTreeClassifier()
```

In [27]:
```
y_pred4=dt.predict(x_test)
```

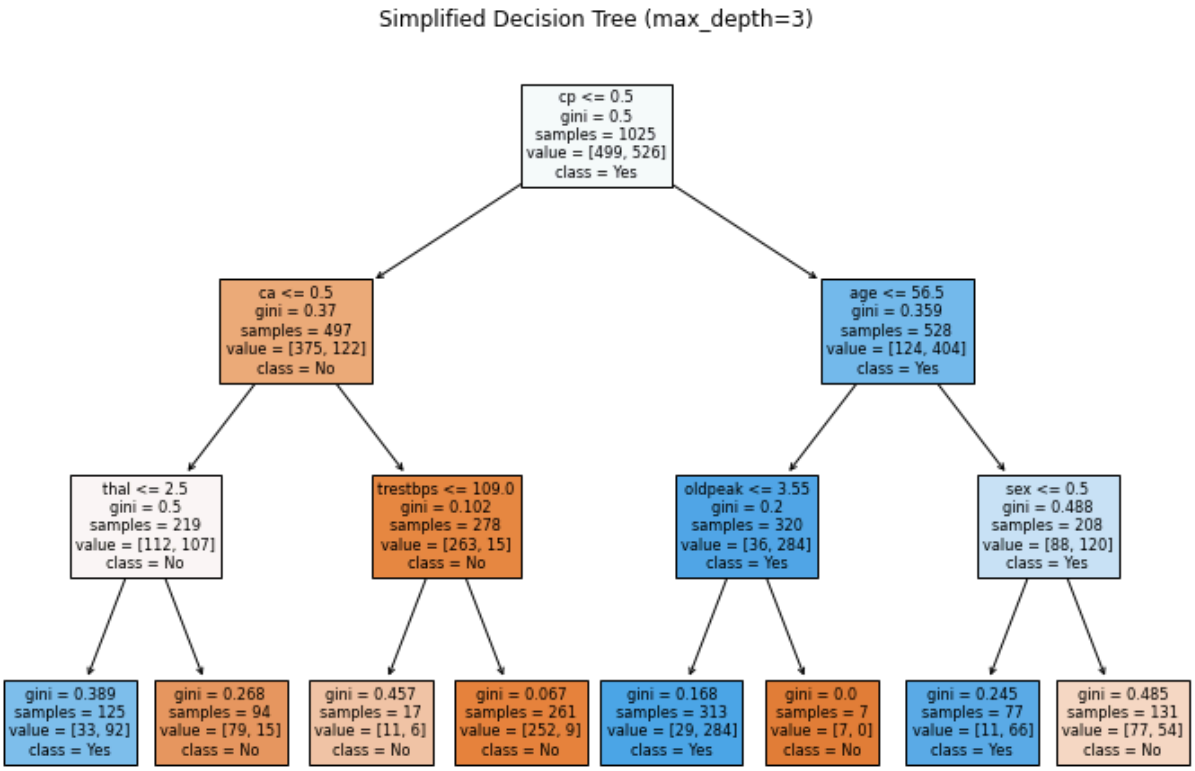
In [28]:
```
accuracy_score (y_test,y_pred4)
```

Out[28]:
```
1.0
```

In [29]:
```
from sklearn.tree import DecisionTreeClassifier, plot_tree
import matplotlib.pyplot as plt

# Create and train the Decision Tree model
model = DecisionTreeClassifier(max_depth=3)  # limit depth
model.fit(x, y)

# Plot the decision tree
plt.figure(figsize=(12, 8))
plot_tree(model, filled=True, feature_names=x.columns, class_names=['No', 'Yes'])
plt.title("Simplified Decision Tree (max_depth=3)")
plt.show()
```



Conclusion :

Theexperiment successfully implemented the Decision Tree algorithm, highlighting its interpretability and effectiveness.

In [ ]:
```
```