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J021

```
In [2]:
import os
print(os.getcwd())
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
C:\Users\User-1
In [3]:
df = pd.read_csv('car_evaluation.csv', header = None)
In [4]:
df.head()
Out[4]:
            1 2 3
                           5
                                 6
0 vhigh vhigh 2 2 small
                          low
                              unacc
1 vhigh vhigh 2 2 small med
                              unacc
2 vhigh vhigh 2 2 small
                         high
                              unacc
3 vhigh vhigh 2 2
                    med
                          low
                              unacc
4 vhigh vhigh 2 2
                    med med unacc
In [5]:
col_names = ['buying', 'maint', 'doors', 'persons', 'lug_boot', 'safety', 'class']
df.columns = col names
col_names
Out[5]:
['buying', 'maint', 'doors', 'persons', 'lug_boot', 'safety', 'class']
```

In [6]:

df.head()

Out[6]:

	buying	maint	doors	persons	lug_boot	safety	class
0	vhigh	vhigh	2	2	small	low	unacc
1	vhigh	vhigh	2	2	small	med	unacc
2	vhigh	vhigh	2	2	small	high	unacc
3	vhigh	vhigh	2	2	med	low	unacc
4	vhigh	vhigh	2	2	med	med	unacc

In [7]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1728 entries, 0 to 1727
Data columns (total 7 columns):
 # Column Non-Null Count Dtype

#	COTUMN	Non-Null Count	Dtype
0	buying	1728 non-null	object
1	maint	1728 non-null	object
2	doors	1728 non-null	object
3	persons	1728 non-null	object
4	lug_boot	1728 non-null	object
5	safety	1728 non-null	object
6	class	1728 non-null	object

dtypes: object(7)
memory usage: 94.6+ KB

```
In [8]:
```

```
for i in col names:
    print(df[i].value_counts())
med
         432
vhigh
         432
high
         432
         432
low
Name: buying, dtype: int64
         432
med
         432
vhigh
high
         432
low
         432
Name: maint, dtype: int64
         432
         432
5more
2
         432
         432
4
Name: doors, dtype: int64
more
        576
2
        576
4
        576
Name: persons, dtype: int64
         576
         576
big
small
         576
Name: lug_boot, dtype: int64
        576
med
high
        576
        576
low
Name: safety, dtype: int64
unacc
         1210
          384
acc
           69
good
           65
vgood
Name: class, dtype: int64
In [9]:
df.shape
Out[9]:
(1728, 7)
In [10]:
X = df.drop(['class'],axis = 1)
y = df['class']
In [11]:
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.3,random_state=42)
```

```
In [12]:
```

```
from sklearn.preprocessing import OrdinalEncoder
enc = OrdinalEncoder()
X_train = enc.fit_transform(X_train)
X_test = enc.transform((X_test))
```

Gini index as criterion

```
In [13]:
from sklearn.tree import DecisionTreeClassifier

In [14]:

clf_gini = DecisionTreeClassifier(criterion='gini', max_depth=3, random_state=42)
clf_gini.fit(X_train, y_train)

Out[14]:
DecisionTreeClassifier(max_depth=3, random_state=42)
In [15]:
```

Grid Search Cv

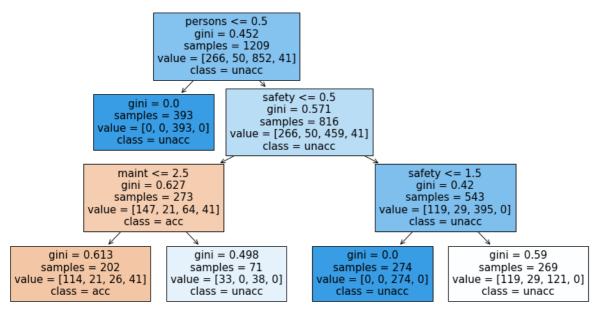
y_pred = clf_gini.predict(X_test)

```
In [48]:
```

```
from sklearn.model_selection import GridSearchCV
option=['gini','entropy']
weight_option=['auto','sqrt','log2']
param_grid = {'criterion': option , 'max_features':[2,3,4,5,6] , 'max_depth':[4,5,6,7] , 'm
grid=GridSearchCV(clf_gini,param_grid,cv=3,scoring='accuracy')
grid.fit(X_train,y_train)
print(grid.best_score_)
print(grid.best_params_)
0.9247311827956989
```

```
{'criterion': 'gini', 'max_depth': 7, 'max_features': 6, 'min_samples_spli
t': 2}
```

In [49]:



In [50]:

```
# Check for underfitting
print(f'Training set score: {clf_gini.score(X_train,y_train)}')
print(f'Test set score: {clf_gini.score(X_test,y_test)}')
```

Training set score: 0.7775020678246485 Test set score: 0.7572254335260116

Model after grid search

In [51]:

```
dtc = DecisionTreeClassifier(criterion='gini', max_depth=7,max_features = 6)
dtc.fit(X_train, y_train)
```

Out[51]:

DecisionTreeClassifier(max_depth=7, max_features=6)

In [61]:

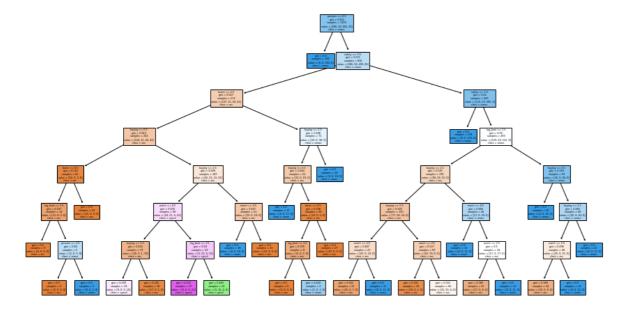
```
y_pred = dtc.predict(X_test)
```

In [53]:

```
print(f'Training set score: {dtc.score(X_train,y_train)}')
print(f'Test set score: {dtc.score(X_test,y_test)}')
```

Training set score: 0.9330024813895782 Test set score: 0.9344894026974951

In [54]:



Cross validation

In [55]:

```
from sklearn.model_selection import cross_val_score
score=cross_val_score(dtc,X_train,y_train,cv=10,scoring='accuracy')
score.mean()
```

Out[55]:

0.920564738292011

In [57]:

```
from sklearn.model_selection import cross_val_score
score=cross_val_score(dtc,X_test,y_test,cv=10,scoring='accuracy')
score.mean()
```

Out[57]:

0.8978883861236803

In [62]:

```
from sklearn.metrics import confusion_matrix, classification_report
cm = confusion_matrix(y_test, y_pred)
```

In [63]:

print(cm)

In [64]:

print(classification_report(y_test, y_pred))

	precision	recall	f1-score	support
acc good	0.84 0.60	0.92 0.32	0.88 0.41	118 19
unacc	1.00	0.97	0.98	358
vgood	0.75	1.00	0.86	24
accuracy			0.93	519
macro avg	0.80	0.80	0.78	519
weighted avg	0.94	0.93	0.93	519

In []: