In [27]:	<pre>import os print(os.getcwd()) import pandas as pd</pre>
	<pre>import numpy as np import matplotlib.pyplot as plt %matplotlib inline</pre>
In [28]:	<pre>C:\Users\HP\Desktop\college\sem 5\ML\car_evaluation.csv', header = None)</pre>
In [29]: Out[29]:	<pre>df.head() 0  1 2 3  4  5  6</pre>
040[20].	<ul> <li>vhigh vhigh 2 2 small low unacc</li> <li>vhigh vhigh 2 2 small med unacc</li> </ul>
	<ul> <li>vhigh vhigh 2 2 small high unacc</li> <li>vhigh vhigh 2 2 med low unacc</li> <li>vhigh vhigh 2 2 med med unacc</li> </ul>
In [30]:	<pre>col_names = ['buying', 'maint', 'doors', 'persons', 'lug_boot', 'safety', 'class'] df.columns = col_names</pre>
Out[30]:	col_names ['buying', 'maint', 'doors', 'persons', 'lug_boot', 'safety', 'class']
In [31]: Out[31]:	ur meau()
out[oi].	0     vhigh     vhigh     2     2     small     low     unacc       1     vhigh     vhigh     2     2     small     med     unacc
	<ul> <li>vhigh vhigh 2 2 small high unacc</li> <li>vhigh vhigh 2 2 med low unacc</li> <li>vhigh vhigh 2 2 med med unacc</li> </ul>
In [32]:	<pre>df.info()</pre>
	<pre><class 'pandas.core.frame.dataframe'=""> RangeIndex: 1728 entries, 0 to 1727 Data columns (total 7 columns): # Column Non-Null Count Dtype</class></pre>
	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)
In [33]:	<pre>memory usage: 94.6+ KB  for i in col_names:     print(df[i].value_counts())</pre>
	vhigh 432 low 432 med 432 high 432
	high 432 Name: buying, dtype: int64 vhigh 432 low 432 med 432
	high 432 Name: maint, dtype: int64 4 432 2 432
	5more 432 3 432 Name: doors, dtype: int64 4 576 2 576
	more 576 Name: persons, dtype: int64 big 576 small 576 med 576
	Name: lug_boot, dtype: int64 low 576 med 576 high 576
	Name: safety, dtype: int64 unacc 1210 acc 384 good 69 vgood 65
In [34]:	Name: class, dtype: int64  df.shape
Out[34]:	(1728, 7)  X = df.drop(['class'].axis = 1)
In [36]:	<pre>X = df.drop(['class'], axis = 1) y = df['class']  from sklearn model selection import train test split</pre>
In [36]:	<pre>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)</pre> from sklearn preprocessing import OrdinalEncoder
±п [37]:	<pre>from sklearn.preprocessing import OrdinalEncoder enc = OrdinalEncoder() X_train = enc.fit_transform(X_train) X_test = enc.transform((X_test))</pre>
	Gini index as criterion
In [38]: In [39]:	<pre>from sklearn.tree import DecisionTreeClassifier</pre>
Out[39]:	<pre>clf_gini = DecisionTreeClassifier(criterion='gini', max_depth=3, random_state=42) clf_gini.fit(X_train, y_train)  DecisionTreeClassifier(max_depth=3, random_state=42)</pre>
In [40]:	<pre>y_pred = clf_gini.predict(X_test)</pre>
In [41]:	Grid Search CV
111 [41].	<pre>from sklearn.model_selection import GridSearchCV  option = ['gini', 'entropy'] weight_options = ['auto', 'sqrt', 'log2'] param_grid = {'criterion': option, 'max_features': [2,3,4,5,6], 'max_depth': [4,5,6,7], 'min_samples_split': [2,3,4,5]} grid = GridSearchCV(clf_gini, param_grid, cv=3, scoring = 'accuracy')</pre>
	<pre>grid = Gridsearchev(cir_gini, param_grid, ev=s, scoring = accuracy) grid.fit(X_train, y_train)  print(grid.best_score_) print(grid.best_params_)</pre>
	0.9247311827956989 {'criterion': 'gini', 'max_depth': 7, 'max_features': 6, 'min_samples_split': 2}
In [42]:	<pre>from sklearn import tree plt.figure(figsize=(15,8)) tree.plot_tree(clf_gini,</pre>
	class_names= list(set(y_train)),     filled = True)  plt.show()
	persons <= 0.5 gini = 0.452
	samples = 1209 value = [266, 50, 852, 41] class = good
	gini = 0.0 samples = 393 value = [0, 0, 393, 0]  safety <= 0.5 gini = 0.571 samples = 816 value = [266, 50, 459, 41]
	class = good
	gini = 0.627 samples = 273 value = [147, 21, 64, 41] class = vgood  gini = 0.42 samples = 543 value = [119, 29, 395, 0] class = good
	gini = 0.613 samples = 202  gini = 0.498 samples = 271  gini = 0.0 samples = 274  gini = 0.59 samples = 269
	value = [114, 21, 26, 41]       value = [33, 0, 38, 0]       value = [0, 0, 274, 0]       value = [119, 29, 121, 0]         class = good       class = good       class = good
In [43]:	<pre># Check for underfitting  print(f'Training set score: {clf_gini.score(X_train,y_train)}')  print(f'Training set score; {clf_gini.score(X_train,y_train)}')</pre>
	<pre>print(f'Test set score: {clf_gini.score(X_test,y_test)}')  Training set score: 0.7775020678246485 Test set score: 0.7572254335260116</pre>
In [46]:	Model after we do GridSearch
Out[46]:	<pre>dtc = DecisionTreeClassifier(criterion = 'gini', max_depth= 7, max_features= 6, min_samples_split= 2) dtc.fit(X_train, y_train)  DecisionTreeClassifier(max_depth=7, max_features=6)</pre>
In [47]:	<pre>y_pred = dtc.predict(X_test)</pre>
	<pre>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</pre>
In [48]:	<pre>from sklearn import tree plt.figure(figsize=(15,8))</pre>
	<pre>tree.plot_tree(dtc,</pre>
	plt.show()
	Image: 12   Image: 13   Imag
	### C 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
	## 1
	Cross Validation
In [50]:	# For Train  from sklearn.model_selection import cross_val_score
0.04.5	<pre>score = cross_val_score(dtc, X_train, y_train, cv=10, scoring='accuracy') score.mean()  0.920564738292011</pre>
Out[50]: In [51]:	# For Test
	<pre>from sklearn.model_selection import cross_val_score  score = cross_val_score(dtc, X_test, y_test, cv=10, scoring='accuracy') score.mean()</pre>
Out[51]: In [53]:	0.8959653092006032
±п [53]:	<pre>from sklearn.metrics import confusion_matrix, classification_report  cm = confusion_matrix(y_test, y_pred)  print(cm)</pre>
	print(cm)  [[109
In [54]:	<pre>[ 11  0 346  1] [ 0  0  0 24]]  print(classification_report(y_test, y_pred))</pre>
	precision recall f1-score support acc 0.84 0.92 0.88 118
	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 [ ]:	