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         J036
 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
                              4
          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
              vhigh
                   vhigh
                                         small
                                                low unacc
                   vhigh
                            2
          1
              vhigh
                                   2
                                         small
                                               med unacc
              vhigh
                   vhigh
                                   2
                                         small
                                               high unacc
                                                low unacc
              vhigh
                   vhigh
                            2
                                   2
                                         med
              vhigh vhigh
                                         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
                         -----
                                          object
          0
              buying
                         1728 non-null
              maint
                         1728 non-null
                                          object
          1
          2
               doors
                         1728 non-null
                                          object
               persons
                         1728 non-null
                                          object
              lug_boot 1728 non-null
                                          object
                         1728 non-null
               safety
                                          object
                         1728 non-null
                                          object
               class
         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
         low
                   432
         Name: buying, dtype: int64
         med
                   432
         vhigh
                   432
         high
                   432
                   432
         low
         Name: maint, dtype: int64
                   432
         5more
                   432
         2
                   432
         4
                   432
         Name: doors, dtype: int64
         more
                  576
         2
                  576
                  576
         Name: persons, dtype: int64
         med
                   576
         big
                   576
         small
                   576
         Name: lug_boot, dtype: int64
         med
                  576
         high
                  576
                  576
         low
         Name: safety, dtype: int64
         unacc
                   1210
         acc
                    384
                     69
         good
         vgood
                     65
         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,ra
         ndom_state=42)
In [12]: from sklearn.preprocessing import OrdinalEncoder
          enc = OrdinalEncoder()
         X_train = enc.fit_transform(X_train)
         X_{\text{test}} = \text{enc.transform}((X_{\text{test}}))
         Gini index as criterion
In [13]: from sklearn.tree import DecisionTreeClassifier
In [14]: | clf_gini = DecisionTreeClassifier(criterion='gini', max_depth=3, random_
         clf_gini.fit(X_train, y_train)
Out[14]: DecisionTreeClassifier(max_depth=3, random_state=42)
In [15]: y_pred = clf_gini.predict(X_test)
         Grid Search Cv
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_de
         pth':[4,5,6,7] , 'min_samples_split':[2,3,4,5]}
         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_sp
         lit': 2}
In [49]: from sklearn import tree
          plt.figure(figsize=(15,8))
          tree.plot_tree(clf_gini,
                         feature_names=['buying', 'maint', 'doors', 'persons', 'lu
         g_boot', 'safety'],
                         class_names= list(set(y_train)),
                         filled = True)
          plt.show()
                                 persons <= 0.5
                                  gini = 0.452
                                 samples = 1209
                              value = [266, 50, 852, 41]
                                           safety \leq 0.5
                          gini = 0.0
                                            gini = 0.571
                        samples = 393
                                           samples = 816
                      value = [0, 0, 393, 0]
                                       value = [266, 50, 459, 41]
                                           class = unacc
                        maint \leq 2.5
                                                              safety \leq 1.5
                         gini = 0.627
                                                               gini = 0.42
                                                              samples = 543
                        samples = 273
                                                           value = [119, 29, 395, 0]
                     value = [147, 21, 64, 41]
                         class = acc
                                                              class = unacc
                                  gini = 0.498
                                                                         gini = 0.59
                gini = 0.613
                                                      gini = 0.0
              samples = 202
                                  samples = 71
                                                     samples = 274
                                                                       samples = 269
                                                   value = [0, 0, 274, 0]
           value = [114, 21, 26, 41]
                               value = [33, 0, 38, 0]
                                                                    value = [119, 29, 121, 0]
                                                     class = unacc
                class = acc
                                  class = unacc
                                                                        class = unacc
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
         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]: from sklearn import tree
          plt.figure(figsize=(15,8))
         tree.plot_tree(dtc,
                         feature_names=['buying', 'maint', 'doors', 'persons', 'lu
         g_boot', 'safety'],
                         class_names= list(set(y_train)),
                         filled = True)
         plt.show()
         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)
         [[109
                      1
                          4]
                          3]
          [ 10
                  6
                      0
                          1]
            11
                  0 346
             0
                         24]]
                  0
                      0
In [64]: print(classification_report(y_test, y_pred))
                        precision
                                      recall f1-score
                                                          support
                                        0.92
                                                   0.88
                             0.84
                                                              118
                   acc
                  good
                             0.60
                                        0.32
                                                   0.41
                                                               19
                 unacc
                             1.00
                                        0.97
                                                   0.98
                                                              358
                 vgood
                             0.75
                                        1.00
                                                   0.86
                                                               24
                                                   0.93
                                                              519
              accuracy
            macro avg
                             0.80
                                        0.80
                                                   0.78
                                                              519
In [ ]:
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