2.DR-Grid

October 13, 2022

```
[1]: #importing the Libraies
     import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
[2]: # Reading the Dataset
     dataset = pd.read_csv('insurance_pre.csv')
[3]: dataset
[3]:
                                   children smoker
                                                          charges
            age
                     sex
                             bmi
     0
             19
                 female
                          27.900
                                                      16884.92400
                                                yes
     1
                                          1
             18
                   male
                          33.770
                                                 no
                                                       1725.55230
     2
             28
                   male
                          33.000
                                          3
                                                 no
                                                       4449.46200
     3
             33
                          22.705
                                          0
                                                     21984.47061
                   male
                                                 no
     4
             32
                          28.880
                                          0
                                                       3866.85520
                   male
                                                 no
                                                 •••
             50
                          30.970
                                          3
                                                      10600.54830
     1333
                   male
     1334
                 female
                          31.920
                                          0
                                                       2205.98080
             18
                                                 no
     1335
                          36.850
             18
                 female
                                                 no
                                                       1629.83350
     1336
             21
                 female
                          25,800
                                          0
                                                       2007.94500
                                                 no
     1337
                 female
                          29.070
                                                     29141.36030
             61
                                                yes
     [1338 rows x 6 columns]
[4]:
    dataset
[4]:
                                   children smoker
                                                          charges
            age
                     sex
                             bmi
     0
             19
                 female
                          27.900
                                          0
                                                yes
                                                      16884.92400
                          33.770
     1
             18
                   male
                                          1
                                                       1725.55230
                                                 no
     2
             28
                          33.000
                                          3
                                                       4449.46200
                   male
                                                 no
     3
             33
                   male
                          22.705
                                          0
                                                     21984.47061
                                                 no
     4
             32
                                          0
                          28.880
                                                       3866.85520
                   male
                                                 no
                                                 •••
     1333
             50
                   male
                          30.970
                                          3
                                                      10600.54830
                                                 no
     1334
                          31.920
                                          0
                                                       2205.98080
             18
                 female
                                                 no
     1335
                 female
                          36.850
                                                 no
                                                       1629.83350
```

```
1337
             61 female 29.070
                                         0
                                              yes 29141.36030
      [1338 rows x 6 columns]
 [5]: dataset=pd.get_dummies(dataset,drop_first=True)
 [6]: dataset
 [6]:
                                        charges sex_male
            age
                    bmi
                         children
                                                           smoker yes
      0
             19 27.900
                                 0 16884.92400
                                                        0
                                                                     1
      1
             18 33.770
                                     1725.55230
                                                        1
                                                                     0
             28 33.000
                                     4449.46200
                                                        1
                                                                     0
                                 0 21984.47061
      3
             33 22.705
                                                        1
                                                                     0
      4
             32 28.880
                                     3866.85520
                                                        1
                                                                     0
             50 30.970
                                 3 10600.54830
      1333
                                                                     0
                                                        1
      1334
             18 31.920
                                     2205.98080
                                                        0
                                                                     0
      1335
             18 36.850
                                   1629.83350
                                                        0
                                                                     0
      1336
             21 25.800
                                     2007.94500
                                                        0
                                                                     0
      1337
             61 29.070
                                 0 29141.36030
                                                                     1
      [1338 rows x 6 columns]
 [7]: indep=dataset[['age', 'bmi', 'children', 'sex_male', 'smoker_yes']]
      dep=dataset['charges']
 [8]: #split into training set and test
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(indep, dep, test_size = 1/
       \rightarrow3, random_state = 0)
 [9]: from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      X_train = sc.fit_transform(X_train)
      X_test = sc.transform(X_test)
 []:
[10]: from sklearn.model_selection import GridSearchCV
      from sklearn.tree import DecisionTreeRegressor
      param_grid = {'criterion':['mse', 'mae', 'friedman_mse'],
                    'max_features': ['auto', 'sqrt', 'log2'],
                    'splitter':['best','random']}
```

0

no

2007.94500

1336

21 female 25.800

```
\rightarrow= 3,n_jobs=-1)
      # fitting the model for grid search
      grid.fit(X_train, y_train)
     C:\Anaconda3\envs\ML\lib\site-packages\sklearn\model_selection\_split.py:1978:
     FutureWarning: The default value of cv will change from 3 to 5 in version 0.22.
     Specify it explicitly to silence this warning.
       warnings.warn(CV_WARNING, FutureWarning)
     [Parallel(n jobs=-1)]: Using backend LokyBackend with 12 concurrent workers.
     Fitting 3 folds for each of 18 candidates, totalling 54 fits
     [Parallel(n_jobs=-1)]: Done 8 tasks
                                                 | elapsed:
                                                               4.5s
     [Parallel(n_jobs=-1)]: Done 50 out of 54 | elapsed:
                                                               4.8s remaining:
                                                                                  0.3s
     [Parallel(n_jobs=-1)]: Done 54 out of 54 | elapsed:
                                                               4.8s finished
     C:\Anaconda3\envs\ML\lib\site-packages\sklearn\model_selection\_search.py:814:
     DeprecationWarning: The default of the `iid` parameter will change from True to
     False in version 0.22 and will be removed in 0.24. This will change numeric
     results when test-set sizes are unequal.
       DeprecationWarning)
[10]: GridSearchCV(cv='warn', error_score='raise-deprecating',
                   estimator=DecisionTreeRegressor(criterion='mse', max_depth=None,
                                                   max_features=None,
                                                   max_leaf_nodes=None,
                                                   min_impurity_decrease=0.0,
                                                   min_impurity_split=None,
                                                   min_samples_leaf=1,
                                                   min samples split=2,
                                                   min_weight_fraction_leaf=0.0,
                                                   presort=False, random_state=None,
                                                   splitter='best'),
                   iid='warn', n_jobs=-1,
                   param_grid={'criterion': ['mse', 'mae', 'friedman_mse'],
                               'max_features': ['auto', 'sqrt', 'log2'],
                               'splitter': ['best', 'random']},
                   pre_dispatch='2*n_jobs', refit=True, return_train_score=False,
                   scoring=None, verbose=3)
[11]: # print best parameter after tuning
      #print(qrid.best_params_)
      re=grid.cv_results_
      #print(re)
      grid_predictions = grid.predict(X_test)
```

grid = GridSearchCV(DecisionTreeRegressor(), param grid, refit = True, verbose

```
# print classification report
      from sklearn.metrics import r2_score
      r_score=r2_score(y_test,grid_predictions)
      print("The R_score value for best parameter {}:".format(grid.
       →best_params_),r_score)
     {'criterion': 'mae', 'max_features': 'auto', 'splitter': 'random'}
     The R_score value for best parameter {'criterion': 'mae', 'max_features':
      'auto', 'splitter': 'random'}: 0.7150762213240649
[13]: table=pd.DataFrame.from_dict(re)
[14]:
     table
[14]:
          mean fit time
                          std fit time
                                        mean_score_time
                                                          std score time
      0
               0.005985
                          1.409110e-03
                                                0.002601
                                                                 0.000353
      1
               0.004910
                          1.115174e-03
                                                0.001689
                                                                 0.000697
      2
                          4.484425e-05
                                                                 0.000000
               0.003957
                                                0.001995
      3
               0.004290 4.253465e-04
                                                0.002125
                                                                 0.000184
      4
               0.003149
                         2.514277e-04
                                                0.001849
                                                                 0.000643
      5
               0.002993 5.947204e-07
                                                0.001995
                                                                 0.000002
      6
                         1.138618e-03
               0.032343
                                                0.001995
                                                                 0.00001
      7
               0.020599
                         2.330010e-03
                                                                0.000823
                                                0.002158
      8
               0.014959 8.157561e-04
                                                0.001798
                                                                 0.000639
      9
               0.011982 1.267615e-03
                                                0.002923
                                                                 0.001311
      10
               0.017405 2.710370e-03
                                                0.001922
                                                                 0.000101
      11
               0.016491 2.369476e-03
                                                0.001588
                                                                 0.000493
      12
               0.006316 9.395378e-04
                                                                 0.000408
                                                0.002508
      13
               0.004315 9.361913e-04
                                                0.002187
                                                                 0.000269
      14
               0.004615 1.638069e-03
                                                0.002187
                                                                 0.000266
                                                                 0.000300
      15
               0.003112 1.718468e-04
                                                0.001781
      16
               0.004992
                          2.157333e-03
                                                0.002986
                                                                 0.001414
      17
               0.003323 4.677745e-04
                                                0.001542
                                                                 0.000414
         param_criterion param_max_features param_splitter
      0
                                        auto
                                                        best
                     mse
      1
                      mse
                                        auto
                                                      random
      2
                                        sqrt
                                                        best
                     mse
      3
                                        sqrt
                                                      random
                     mse
      4
                                                        best
                     mse
                                        log2
      5
                     mse
                                        log2
                                                      random
      6
                     mae
                                        auto
                                                        best.
      7
                                        auto
                                                      random
                     mae
      8
                                        sqrt
                                                        best
                     mae
      9
                                        sqrt
                                                      random
                     mae
      10
                      mae
                                        log2
                                                        best
```

```
11
                                   log2
                                                 random
                mae
12
      friedman mse
                                   auto
                                                   best
13
      friedman_mse
                                   auto
                                                 random
14
      friedman_mse
                                                   best
                                   sqrt
15
      friedman_mse
                                   sqrt
                                                 random
16
      friedman_mse
                                                   best
                                   log2
17
      friedman_mse
                                   log2
                                                 random
                                                          split0_test_score
                                                  params
    {'criterion': 'mse', 'max features': 'auto', '...
                                                                  0.538684
0
    {'criterion': 'mse', 'max_features': 'auto', '...
1
                                                                  0.594686
2
    {'criterion': 'mse', 'max_features': 'sqrt', '...
                                                                  0.679781
    {'criterion': 'mse', 'max_features': 'sqrt', '...
3
                                                                  0.691520
4
    {'criterion': 'mse', 'max_features': 'log2', '...
                                                                  0.529707
    {'criterion': 'mse', 'max features': 'log2', '...
5
                                                                  0.554397
6
    {'criterion': 'mae', 'max_features': 'auto',
                                                                  0.647916
7
    {'criterion': 'mae', 'max_features': 'auto', '...
                                                                  0.616340
8
    {'criterion': 'mae', 'max_features': 'sqrt',
                                                                  0.435070
9
    {'criterion': 'mae', 'max_features': 'sqrt', '...
                                                                  0.725520
   {'criterion': 'mae', 'max_features': 'log2', '...
10
                                                                  0.567610
    {'criterion': 'mae', 'max_features': 'log2', '...
11
                                                                  0.532352
    {'criterion': 'friedman_mse', 'max_features': ...
12
                                                                  0.509355
   {'criterion': 'friedman_mse', 'max_features': ...
13
                                                                  0.580917
   {'criterion': 'friedman mse', 'max features': ...
14
                                                                  0.523019
    {'criterion': 'friedman_mse', 'max_features': ...
15
                                                                  0.598454
    {'criterion': 'friedman mse', 'max features': ...
                                                                  0.659411
    {'criterion': 'friedman_mse', 'max_features': ...
                                                                  0.565314
    split1_test_score
                        split2_test_score
                                            mean_test_score
                                                               std_test_score
0
             0.677963
                                  0.694056
                                                    0.636791
                                                                     0.069798
1
             0.657129
                                  0.675125
                                                    0.642260
                                                                     0.034487
2
             0.680816
                                                    0.667265
                                  0.641157
                                                                     0.018451
3
             0.644275
                                  0.591872
                                                    0.642610
                                                                     0.040710
4
             0.540047
                                  0.683060
                                                    0.584210
                                                                     0.069966
5
             0.595953
                                  0.555577
                                                    0.568626
                                                                     0.019313
6
             0.652077
                                  0.688474
                                                    0.662806
                                                                     0.018214
7
             0.707618
                                  0.695565
                                                    0.673111
                                                                     0.040510
8
             0.594457
                                  0.685666
                                                    0.571578
                                                                     0.103603
9
             0.359665
                                  0.504119
                                                    0.529987
                                                                     0.150515
10
             0.310512
                                  0.495136
                                                    0.457876
                                                                     0.108239
11
             0.573472
                                  0.664373
                                                    0.590001
                                                                     0.055162
                                                                     0.081557
12
             0.671252
                                  0.691500
                                                    0.623907
13
             0.630293
                                  0.631976
                                                    0.614358
                                                                     0.023696
14
             0.671363
                                  0.676148
                                                    0.623397
                                                                     0.071124
15
             0.535226
                                  0.656763
                                                    0.596816
                                                                     0.049603
16
             0.621462
                                  0.720891
                                                    0.667246
                                                                     0.040946
17
             0.439232
                                  0.515024
                                                    0.506589
                                                                     0.051831
```

```
0
                         7
                         6
      1
                         2
      2
                         5
      3
      4
                        13
      5
                        15
      6
                         4
      7
                         1
      8
                        14
      9
                        16
                        18
      10
                        12
      11
      12
                         8
      13
                        10
      14
                         9
      15
                        11
      16
                         3
      17
                        17
[15]: age_input=float(input("Age:"))
      bmi_input=float(input("BMI:"))
      children_input=float(input("Children:"))
      sex_male_input=int(input("Sex Male 0 or 1:"))
      smoker_yes_input=int(input("Smoker Yes 0 or 1:"))
     Age:4
     BMI:56
     Children:5
     Sex Male 0 or 1:5
     Smoker Yes 0 or 1:3
[16]: Future_Prediction=grid.
       →predict([[age_input,bmi_input,children_input,sex_male_input,smoker_yes_input]])#_
       \hookrightarrow change the paramter, play with it.
      print("Future_Prediction={}".format(Future_Prediction))
     Future_Prediction=[49577.6624]
 []:
 []:
 []:
 []:
```

rank_test_score

[]:	
[]:	