DT on boston dataset

April 23, 2021

```
[1]: import numpy as np
    import pandas as pd
    from sklearn.datasets import load_boston
    from sklearn.metrics import mean_squared_error
    import random
    from sklearn.tree import DecisionTreeRegressor
    from prettytable import PrettyTable
[2]: boston = load_boston()
    x=boston.data
    y=boston.target
[3]: df = pd.DataFrame(x, columns=boston.feature_names)
    df['PRICE'] = v
    no_of_total_columns = len(df.columns) - 1
    df.head()
[3]:
           CRIM
                  ZN
                      INDUS CHAS
                                     NOX
                                             RM
                                                  AGE
                                                          DIS
                                                               RAD
                                                                      TAX \
    0 0.00632 18.0
                       2.31
                              0.0 0.538
                                          6.575
                                                 65.2 4.0900
                                                               1.0
                                                                    296.0
                       7.07
    1 0.02731
                 0.0
                              0.0 0.469
                                          6.421
                                                78.9 4.9671
                                                               2.0
                                                                    242.0
    2 0.02729
                 0.0
                       7.07
                              0.0 0.469
                                          7.185 61.1 4.9671
                                                               2.0 242.0
    3 0.03237
                 0.0
                       2.18
                              0.0 0.458
                                          6.998 45.8 6.0622
                                                              3.0
                                                                    222.0
    4 0.06905
                 0.0
                       2.18
                              0.0 0.458 7.147 54.2 6.0622 3.0 222.0
       PTRATIO
                     B LSTAT PRICE
                         4.98
    0
          15.3
                396.90
                                24.0
           17.8 396.90
                         9.14
                                21.6
    1
    2
           17.8 392.83
                         4.03
                                34.7
    3
           18.7
                394.63
                         2.94
                                33.4
    4
           18.7 396.90
                         5.33
                                36.2
[4]: def generating_samples(x, y):
         # random rows 303, keep it
         selecting_rows = sorted(np.random.choice(a=506, size=303, replace=False),__
      →reverse=False)
```

```
# randomly choose columns 3<= col <=13
         rnd_number = random.randint(3, 13)
         selecting_cols = sorted(np.random.choice(a=13, size=rnd_number,__
      →replace=False), reverse=False)
         # Preparing Data
         # selecting_row_ and selecting_cols data
         sample_data = x[selecting_rows,:]
         sample_data = sample_data[:, selecting_cols]
         sample_target = y[selecting_rows]
         # replaced row data
         # random 203 points from 303
         replacing_rows = sorted(np.random.choice(a=303, size=203, replace=False),_
      →reverse=False)
         replaced_data = sample_data[replacing_rows]
         replaced_target = sample_target[replacing_rows]
         # stacking selected_data and replaced_data
         final_sample_data = np.vstack((sample_data, replaced_data))
         final_sample_target = np.vstack((sample_target.reshape(-1, 1),__
      →replaced_target.reshape(-1, 1)))
         # return sampled data_points, sampled target data, sampled_row indexs,_
      \rightarrowsampled column index
         return final sample data, final sample target, selecting rows, u
      →selecting_cols
[5]: final_sample_data, final_sample_target, selecting_rows, selecting_cols =

→generating_samples(x, y)
[6]: def grader_samples(a,b,c,d):
         length = (len(a) == 506 and len(b) == 506)
         sampled = (len(a)-len(set([str(i) for i in a]))==203)
         rows_length = (len(c) == 303)
```

assert(length and sampled and rows_length and column_length)

column_length= (len(d)>=3)

```
return True
a,b,c,d = generating_samples(x, y)
grader_samples(a,b,c,d)
```

[6]: True

1 Task -1 (Creating 30 samples)

[7]: True

2 Storing 30 models

```
[8]: list_of_models = []

for input_data, output_data in zip(list_of_input_data, list_of_output_data):
    model = DecisionTreeRegressor(max_depth=None)
    model.fit(input_data, output_data)
    list_of_models.append(model)
```

```
print(len(list_of_models))
```

30

3 Calculating MSE

[10]: 2.3622332015810272

4 Calculating OOB Score

```
# do calculataion and push to oob_y_pred
y_pred = np.median(np.asarray(temp_y_pred))

oob_y_pred.append(y_pred)

oob_y_pred = np.array(oob_y_pred).reshape(-1, 1)

print(oob_y_pred.shape)
print(y.shape)
print("true: {}, pred: {}".format(y[0], oob_y_pred[0]))

(506, 1)
(506, 1)
true: [24.], pred: [31.95]

[21]: oob_score = np.square(np.subtract(oob_y_pred, y)).mean()
oob_score
```

[21]: 14.016645223030086

5 Train MSE Scores

```
for i in range(35):
    # sampling

list_of_input_data = []
    list_of_selected_rows = []
    list_of_selected_cols = []

for i in range(30):

    final_sample_data, final_sample_target, selecting_rows, selecting_cols_
    = generating_samples(x, y)

list_of_input_data.append(final_sample_data)
    list_of_output_data.append(final_sample_target)
    list_of_selected_rows.append(selecting_rows)
    list_of_selected_cols.append(selecting_cols)
```

```
# List of models
          list_of_models = []
          for input_data, output_data in zip(list_of_input_data, list_of_output_data):
              model = DecisionTreeRegressor(max_depth=None)
              model.fit(input data, output data)
              list_of_models.append(model)
          # Predicting and calculating MSE
          y_pred_all_models = []
          for col_sampling, model in zip(list_of_selected_cols, list_of_models):
              x_input = x[:, col_sampling]
              y_pred_all_models.append(model.predict(x_input).reshape(-1, 1))
          y_pred = np.around(np.divide(np.sum(y_pred_all_models, axis=0), 30), 1)
          mse = np.square(np.subtract(y, y_pred)).mean()
          #Stored in a list
          train_mse_scores.append(mse)
[14]: print(len(train_mse_scores))
      print(train_mse_scores)
     35
     [2.8683399209486167, 2.155790513833992, 2.7674901185770744, 2.1426086956521737,
     2.1098616600790514, 2.792351778656126, 2.1528063241106716, 2.4678260869565216,
     1.9978063241106718, 2.030098814229249, 2.4497628458498024, 2.873458498023715,
     2.2759486166007896, 2.1954150197628457, 2.056383399209486, 3.0770750988142295,
```

2.886324110671936, 2.5661462450592882, 2.1811264822134384, 1.9558102766798422, 1.903102766798419, 1.975039525691699, 2.7585770750988137, 2.058300395256917, 2.3936363636363, 2.2669367588932805, 2.1659486166007906, 2.1801778656126483, 2.181916996047431, 2.1927075098814233, 2.4847430830039525, 2.365415019762845,

2.1672529644268774, 2.9212845849802362, 2.144802371541502]

6 OOB Scores

```
[29]: test_oob_scores = []
      for i in range(35):
          # sampling
          list_of_input_data = []
          list_of_output_data = []
          list_of_selected_rows = []
          list_of_selected_cols = []
          for i in range(30):
              final_sample_data, final_sample_target, selecting_rows, selecting_cols_
       ⇒= generating_samples(x, y)
              list_of_input_data.append(final_sample_data)
              list_of_output_data.append(final_sample_target)
              list_of_selected_rows.append(selecting_rows)
              list_of_selected_cols.append(selecting_cols)
          # List of models
          list_of_models = []
          for input_data, output_data in zip(list_of_input_data, list_of_output_data):
              model = DecisionTreeRegressor(max_depth=None)
              model.fit(input_data, output_data)
              list_of_models.append(model)
          # Predicting and calculating OOB
          oob_y_pred = []
          for data_index, data_row in enumerate(x):
              temp_y_pred = []
              for selected_rows_index, selected_col_index, model in_
       -zip(list_of_selected_rows, list_of_selected_cols, list_of_models):
                  if data_index not in selected_rows_index:
                      # do column sampling
```

```
x_input = np.array(data_row).reshape(1, -1)[:,__
selected_col_index]

# predict push to temp_y_pred
temp_y_pred.append(model.predict(x_input))

# do calculatation and push to oob_y_pred
y_pred = np.median(np.asarray(temp_y_pred))

oob_y_pred.append(y_pred)

oob_y_pred = np.array(oob_y_pred).reshape(-1, 1)

oob_score = np.square(np.subtract(oob_y_pred, y)).mean()

#Stored in a list
test_oob_scores.append(oob_score)
```

```
[30]: print(len(test_oob_scores))
print(test_oob_scores)
```

```
35
[14.337240146906693, 14.787600217508244, 13.594189723320158, 14.80823740118577, 14.723413916564132, 14.007485704874835, 13.388359683794466, 17.450867988871654, 15.38008166431686, 12.239359295277827, 14.263373164683083, 12.384036807740223, 18.182400221849957, 13.205228782290394, 14.705339117994498, 13.74955533596838, 14.562552170820874, 15.344541749011858, 14.014589787474696, 14.65601909012268, 15.006899436383577, 11.434801129425535, 14.394506779924697, 16.228079161176986, 13.42788098375055, 12.524988123124661, 12.155026124776606, 13.67620937637242, 15.818227903926172, 14.148537989529608, 12.501847277119015, 19.897433064891807, 14.624507707509881, 13.771521739130433, 13.860377327115085]
```

7 Computing CI of OOB Score and Train MSE

```
[31]: x = PrettyTable()
x = PrettyTable(["#samples", "Sample Size", "MSE", "Left C.I MSE", "Right C.I

→MSE", "MSE Catch", "OOB", "Left C.I OOB", "Right C.I OOB", "OOB Catch"])

count = 0
```

```
for mse, oob in zip(train_mse_scores, test_oob_scores):
    mse = np.round(mse, 3)
    oob = np.round(oob, 3)
    standard_error_mse = np.sqrt(np.divide((2*mse), 506))
    ci_mse_left = np.round((mse - standard_error_mse), 3)
    ci_mse_right = np.round((mse + standard_error_mse), 3)
    standard_error_oob = np.sqrt(np.divide((2*oob), 506))
    ci_oob_left = np.round((oob - standard_error_oob), 3)
    ci_oob_right = np.round((oob + standard_error_oob), 3)
    row = []
    row.append(i+1)
    row.append("506")
    row.append(mse)
    row.append(ci_mse_left)
    row.append(ci_mse_right)
    row.append((mse >= ci_mse_left) and (mse <= ci_mse_right))</pre>
    row.append(oob)
    row.append(ci_oob_left)
    row.append(ci_oob_right)
    row.append((oob >= ci_oob_left) and (oob <= ci_oob_right))</pre>
    x.add_row(row)
    count += 1
print(x)
```

```
+-----
----+
| #samples | Sample Size | MSE | Left C.I MSE | Right C.I MSE | MSE Catch |
   | Left C.I OOB | Right C.I OOB | OOB Catch |
30
          506
                | 2.217 |
                        2.123
                                  2.311
                                            True
                                                14.337
       14.099
                 14.575
                           True
                | 2.94 |
                              30
          506
                        2.832
                                  3.048
                                            True
                                                1
14.788
       14.546
                 15.03
                           True
                              30
          506
                | 2.661 |
                        2.558
                                  2.764
                                            True
13.594 l
       13.362
                 13.826
                           True
                        1
                1 2.632 |
                        2.53
                                  2.734
                                            True
                                                Ι
   30
          506
                 15.05
14.808 |
       14.566
              True
```

30	l 506	2.54	2.44	2.64	I	True	ı
14.723			True				
J 30		2.355	2.259		- 1	True	
14.007		14.242	True				
30		2.567			I	True	
13.388		13.618					
30		3.351	3.236		ı	True	
17.451		17.714	True				
30		3.001	2.892		ı	True	ı
		15.627					
30		2.123	2.031		ı	True	l
12.239		12.459	True			_	
30		1.99			ı	True	
14.263		14.5				_	
30		2.436	2.338		ı	True	ı
12.384	12.163		True				
30		2.822			ı	True	
			True			_	
30			2.39		ı	True	ı
13.205	12.977		True			_	
30		2.364			ı	True	ı
14.705		14.946	True			_	
30		2.185	2.092		ı	True	
13.75	13.517		True				
30		2.517			ı	True	
14.563			True			_	
30		2.339			ı	True	ı
15.345			True				
30			2.62		ı	True	ı
14.015		14.25					
30		2.311	2.215		ı	True	ı
	14.415		True				
	506		2.159		ı	True	ı
			True				
30			2.643		ı	True	ı
			True			m	
30		2.246			ı	True	ı
			True			m	
30			1.957		I	True	ı
			True			Т	
30		2.241			I	True	ı
13.428			True			Т	
30		2.277			I	True	ı
			True		1	Т	ı
30		2.621	2.519 True		I	True	I
12.155 30					ı	Т~~~	ı
			2.938		I	rrue	1
13.676	13.444	13.908	True				

```
506
                         | 2.74 |
                                       2.636
                                                       2.844
                                                                      True
     30
15.818 |
            15.568
                            16.068
                                            True
                                       | 3.007 |
                                       2.898
                                                       3.116
                                                                      True
     30
                 506
14.149
            13.913
                            14.385
                                            True
     30
                 506
                         | 2.302 |
                                       2.207
                                                       2.397
                                                                      True
                                                                              Τ
                                                 12.502 |
            12.28
                            12.724
                                            True
     30
                 506
                          | 2.584 |
                                       2.483
                                                       2.685
                                                                      True
19.897 |
            19.617
                            20.177
                                            True
     30
                 506
                         | 2.132 |
                                        2.04
                                                       2.224
                                                                      True
                                                                              14.625 |
            14.385
                            14.865
                                            True
                 506
                          | 2.746 |
                                       2.642
                                                        2.85
                                                                      True
     30
13.772
            13.539
                            14.005
                                            True
                                                       2.094
    30
                 506
                         | 2.005 |
                                       1.916
                                                 True
                            14.094
13.86
            13.626
                                            True
```

8 Task 3 (Predicting a data)

```
[43]: y_pred
```

[43]: array([19.5])