## practical-04

## April 26, 2024

```
[42]: import numpy as np
      import pandas as pd
      from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LinearRegression
      from sklearn.metrics import mean_squared_error, r2_score
 [3]: data = pd.read_csv("C:/Users/gugal/Desktop/THIRD 2/PRACTICALS/DS/CODES/DATASETS/
       →HousingData.csv")
      data
 [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
                                                                            296
      1
           0.02731
                      0.0
                            7.07
                                   0.0
                                        0.469
                                                6.421
                                                       78.9
                                                              4.9671
                                                                        2
                                                                           242
      2
           0.02729
                      0.0
                            7.07
                                   0.0
                                        0.469
                                                7.185
                                                       61.1
                                                              4.9671
                                                                        2
                                                                           242
      3
           0.03237
                            2.18
                                        0.458
                                                                           222
                      0.0
                                   0.0
                                                6.998
                                                       45.8
                                                              6.0622
      4
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                                   0.0
                                        0.458
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                                                              6.0622
                                                                           222
                      0.0
                                                7.147
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           0.06263
                                        0.573
                                                              2.4786
                                                                           273
      501
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      502 0.04527
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                                                6.120
                                                       76.7
                                                              2.2875
                                                                           273
      503 0.06076
                      0.0 11.93
                                        0.573
                                                6.976
                                                       91.0
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                     396.90
                              9.14
                                    21.6
      2
              17.8
                    392.83
                              4.03
                                    34.7
      3
              18.7
                     394.63
                              2.94
                                    33.4
      4
              18.7
                     396.90
                               {\tt NaN}
                                    36.2
      501
              21.0
                                    22.4
                     391.99
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      502
              21.0
                     396.90
                              9.08
                                    20.6
      503
              21.0
                                    23.9
                     396.90
                              5.64
      504
              21.0
                     393.45
                              6.48
                                    22.0
      505
              21.0 396.90
                              7.88 11.9
```

```
data.head()
 [8]:
 [8]:
             CRIM
                     ZN
                          INDUS
                                  CHAS
                                          NOX
                                                   RM
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                                                                                  PTRATIO
         0.00632
                   18.0
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                                   0.0
                                        0.538
                                                6.575
                                                        65.2
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                                                                                     15.3
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                                                6.421
                                                        78.9
                                                              4.9671
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                                                                                     17.8
         0.02729
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                                                              4.9671
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                          21.6
         392.83
                   4.03
                          34.7
      3
         394.63
                   2.94
                          33.4
         396.90
                    NaN
                          36.2
 [4]:
     data.isnull().sum()
 [4]: CRIM
                  20
                  20
      ZN
      INDUS
                  20
      CHAS
                  20
      NOX
                   0
      RM
                   0
      AGE
                  20
      DIS
                   0
      RAD
                   0
      TAX
                   0
      PTRATIO
                   0
      В
                   0
      LSTAT
                  20
      MEDV
                   0
      dtype: int64
[22]:
      data.fillna(data.mean(), inplace=True)
[23]:
      data
[23]:
               CRIM
                        ZN
                            INDUS
                                    CHAS
                                            NOX
                                                     RM
                                                                AGE
                                                                         DIS
                                                                              RAD
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      0
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                             2.31
                                     0.0
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                                                          78.900000
                                                                      4.9671
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                                                  6.421
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                                     0.0
                                          0.469
                                                  7.185
                                                          61.100000
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                                                                                 2
                                                                                    242
      3
            0.03237
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                                     0.0
                                          0.458
                                                  6.998
                                                          45.800000
                                                                      6.0622
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                                                                                    222
      4
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                             2.18
                                                  7.147
                                                                                 3
      501
            0.06263
                       0.0
                            11.93
                                     0.0
                                          0.573
                                                  6.593
                                                          69.100000
                                                                      2.4786
                                                                                 1
                                                                                    273
      502
           0.04527
                       0.0
                            11.93
                                     0.0 0.573
                                                  6.120
                                                          76.700000
                                                                      2.2875
                                                                                 1
                                                                                    273
```

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504 0.10959
                     0.0 11.93
                                  0.0 0.573 6.794
                                                                2.3889
                                                                             273
                                                     89.300000
                                  0.0 0.573 6.030
      505 0.04741
                     0.0 11.93
                                                     68.518519
                                                                2.5050
                                                                             273
          PTRATIO
                        В
                                LSTAT MEDV
              15.3 396.90
                             4.980000
      0
                                       24.0
      1
              17.8 396.90
                             9.140000
                                       21.6
      2
              17.8 392.83
                             4.030000 34.7
      3
              18.7 394.63
                             2.940000 33.4
      4
              18.7 396.90
                           12.715432 36.2
              •••
                              ... ...
      . .
                    •••
      501
              21.0 391.99
                            12.715432 22.4
      502
              21.0 396.90
                             9.080000
                                       20.6
              21.0 396.90
      503
                             5.640000 23.9
      504
              21.0 393.45
                             6.480000 22.0
              21.0 396.90
      505
                             7.880000 11.9
      [506 rows x 14 columns]
[24]: data.isnull().sum()
[24]: CRIM
                 0
     ZN
                 0
      INDUS
                 0
      CHAS
                 0
      NOX
      RM
                 0
      AGE
     DIS
     RAD
                 0
     TAX
                 0
     PTRATIO
                 0
     В
                 0
     LSTAT
                 0
     MEDV
      dtype: int64
[25]: data.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 506 entries, 0 to 505
     Data columns (total 14 columns):
          Column
                   Non-Null Count Dtype
          CRIM
                                   float64
      0
                   506 non-null
                   506 non-null
                                   float64
      1
          ZN
```

0.0 0.573 6.976 91.000000

2.1675

1 273

503 0.06076

INDUS

506 non-null

0.0 11.93

float64

```
506 non-null
3
   CHAS
                             float64
4
   NOX
             506 non-null
                             float64
5
   RM
             506 non-null
                             float64
6
   AGE
             506 non-null
                             float64
7
             506 non-null
   DIS
                             float64
8
   RAD
             506 non-null
                             int64
9
             506 non-null
   TAX
                             int64
10 PTRATIO 506 non-null
                             float64
11 B
             506 non-null
                             float64
12 LSTAT
             506 non-null
                             float64
13 MEDV
             506 non-null
                             float64
```

 ${\tt dtypes: float64(12), int64(2)}$ 

memory usage: 55.5 KB

## [26]: data.columns

[26]: Index(['CRIM', 'ZN', 'INDUS', 'CHAS', 'NOX', 'RM', 'AGE', 'DIS', 'RAD', 'TAX', 'PTRATIO', 'B', 'LSTAT', 'MEDV'], dtype='object')

## [27]: data.describe()

[27]:		CRIM	ZN	INDUS	CHAS	NOX	RM	\
	count	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	
	mean	3.611874	11.211934	11.083992	0.069959	0.554695	6.284634	
	std	8.545770	22.921051	6.699165	0.250233	0.115878	0.702617	
	min	0.006320	0.000000	0.460000	0.000000	0.385000	3.561000	
	25%	0.083235	0.000000	5.190000	0.000000	0.449000	5.885500	
	50%	0.290250	0.000000	9.900000	0.000000	0.538000	6.208500	
	75%	3.611874	11.211934	18.100000	0.000000	0.624000	6.623500	
	max	88.976200	100.000000	27.740000	1.000000	0.871000	8.780000	
		AGE	DIS	RAD	TAX	PTRATIO	В	\
	count	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	
	mean	68.518519	3.795043	9.549407	408.237154	18.455534	356.674032	
	std	27.439466	2.105710	8.707259	168.537116	2.164946	91.294864	
	min	2.900000	1.129600	1.000000	187.000000	12.600000	0.320000	
	25%	45.925000	2.100175	4.000000	279.000000	17.400000	375.377500	
	50%	74.450000	3.207450	5.000000	330.000000	19.050000	391.440000	
	75%	93.575000	5.188425	24.000000	666.000000	20.200000	396.225000	
	max	100.000000	12.126500	24.000000	711.000000	22.000000	396.900000	
		LSTAT	MEDV					
	count	506.000000	506.000000					
	mean	12.715432	22.532806					
	std	7.012739	9.197104					
	min	1.730000	5.000000					

```
7.230000
      25%
                          17.025000
      50%
              11.995000
                          21.200000
      75%
              16.570000
                          25.000000
              37.970000
                          50.000000
      max
[28]: X = data.drop('MEDV', axis=1)
      Y = data['MEDV']
[29]: X.columns
[29]: Index(['CRIM', 'ZN', 'INDUS', 'CHAS', 'NOX', 'RM', 'AGE', 'DIS', 'RAD', 'TAX',
             'PTRATIO', 'B', 'LSTAT'],
            dtype='object')
[30]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2,__
       →random_state=42)
[31]: print(len(Y_test))
     102
[32]: model = LinearRegression()
      model.fit(X_train,Y_train)
[32]: LinearRegression()
[37]: prediction = model.predict(X_test)
[41]: mse = mean_squared_error(Y_test,prediction)
      mse
[41]: 25.017672023842703
[43]: rmse = np.sqrt(mse)
      rmse
[43]: 5.001766890194174
[45]: r2 = r2_score(Y_test, prediction)
      r2
[45]: 0.658852019550814
```