```
In [1]: import pandas as pd
        import matplotlib.pyplot as plt
In [2]: df = pd.read csv('HousingData.csv')
                CRIM
                        ZN INDUS CHAS
                                            NOX
                                                    RM AGE
                                                                  DIS RAD TAX PTRAT
Out[2]:
           0 0.00632
                       18.0
                               2.31
                                                                                       1!
                                       0.0 0.538 6.575
                                                         65.2 4.0900
                                                                          1
                                                                             296
                               7.07
                                                                                       1
           1 0.02731
                        0.0
                                       0.0 0.469 6.421
                                                         78.9 4.9671
                                                                             242
           2 0.02729
                        0.0
                               7.07
                                       0.0 0.469 7.185
                                                         61.1 4.9671
                                                                          2
                                                                             242
                                                                                       1
           3 0.03237
                               2.18
                                       0.0 0.458 6.998
                                                                             222
                        0.0
                                                         45.8 6.0622
                                                                          3
                                                                                       18
           4 0.06905
                        0.0
                               2.18
                                       0.0 0.458 7.147
                                                         54.2 6.0622
                                                                             222
                                                                                       18
                                                                          3
         501 0.06263
                              11.93
                                                                                       2:
                        0.0
                                       0.0 0.573 6.593 69.1 2.4786
                                                                          1
                                                                             273
         502 0.04527
                                                                                       2:
                        0.0
                              11.93
                                       0.0 0.573 6.120
                                                         76.7 2.2875
                                                                             273
         503 0.06076
                        0.0
                              11.93
                                       0.0 0.573 6.976
                                                         91.0 2.1675
                                                                             273
                                                                                       2:
                                                                          1
         504 0.10959
                        0.0
                              11.93
                                       0.0 0.573 6.794
                                                         89.3 2.3889
                                                                             273
                                                                                       2:
                                                                          1
         505 0.04741
                        0.0
                              11.93
                                       0.0 0.573 6.030 NaN 2.5050
                                                                             273
                                                                                       2:
                                                                          1
        506 \text{ rows} \times 14 \text{ columns}
In [3]:
        df.shape
Out[3]: (506, 14)
In [4]: df.isnull().sum()
Out[4]: CRIM
                    20
                    20
         ZN
         INDUS
                    20
         CHAS
                    20
         NOX
                     0
         RM
                     0
                    20
         AGE
         DIS
                     0
         RAD
                     0
         TAX
                     0
         PTRATIO
                     0
         В
                     0
         LSTAT
                    20
         MEDV
                     0
         dtype: int64
In [5]: df['CRIM'] = df['CRIM'].fillna(df['CRIM'].mean())
        df['ZN'] = df['ZN'].fillna(df['ZN'].mean())
```

```
df['INDUS'] = df['INDUS'].fillna(df['INDUS'].mean())
df['CHAS'] = df['CHAS'].fillna(df['CHAS'].mean())
df['AGE'] = df['AGE'].fillna(df['NOX'].mean())
df['LSTAT'] = df['LSTAT'].fillna(df['LSTAT'].mean())
df
```

Out[5]:		CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	I
	0	0.00632	18.0	2.31	0.0	0.538	6.575	65.200000	4.0900	1	296	
	1	0.02731	0.0	7.07	0.0	0.469	6.421	78.900000	4.9671	2	242	
	2	0.02729	0.0	7.07	0.0	0.469	7.185	61.100000	4.9671	2	242	
	3	0.03237	0.0	2.18	0.0	0.458	6.998	45.800000	6.0622	3	222	
	4	0.06905	0.0	2.18	0.0	0.458	7.147	54.200000	6.0622	3	222	
	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	
	503	0.06076	0.0	11.93	0.0	0.573	6.976	91.000000	2.1675	1	273	
	504	0.10959	0.0	11.93	0.0	0.573	6.794	89.300000	2.3889	1	273	
	505	0.04741	0.0	11.93	0.0	0.573	6.030	0.554695	2.5050	1	273	

506 rows \times 14 columns

```
In [6]: df.isnull().sum()
Out[6]: CRIM
                     0
         \mathsf{ZN}
                     0
         INDUS
                     0
         CHAS
                     0
         NOX
                     0
         RM
                     0
         AGE
                     0
         DIS
                     0
         RAD
         TAX
                     0
         PTRATIO
                     0
                     0
         LSTAT
                     0
         MEDV
         dtype: int64
In [7]: x = df.drop('MEDV', axis=1)
        Χ
```

Out[7]:		CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	I
	0	0.00632	18.0	2.31	0.0	0.538	6.575	65.200000	4.0900	1	296	
	1	0.02731	0.0	7.07	0.0	0.469	6.421	78.900000	4.9671	2	242	
	2	0.02729	0.0	7.07	0.0	0.469	7.185	61.100000	4.9671	2	242	
	3	0.03237	0.0	2.18	0.0	0.458	6.998	45.800000	6.0622	3	222	
	4	0.06905	0.0	2.18	0.0	0.458	7.147	54.200000	6.0622	3	222	
	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	
	503	0.06076	0.0	11.93	0.0	0.573	6.976	91.000000	2.1675	1	273	
	504	0.10959	0.0	11.93	0.0	0.573	6.794	89.300000	2.3889	1	273	
	505	0.04741	0.0	11.93	0.0	0.573	6.030	0.554695	2.5050	1	273	

 $506 \text{ rows} \times 13 \text{ columns}$

```
In [8]: y = df['MEDV']
         У
 Out[8]: 0
                 24.0
                 21.6
          1
          2
                 34.7
          3
                 33.4
                 36.2
          501
                 22.4
          502
                 20.6
          503
                 23.9
          504
                 22.0
          505
                 11.9
         Name: MEDV, Length: 506, dtype: float64
 In [9]: from sklearn.model_selection import train_test_split
         x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, rar
In [10]: x_train
```

Out[10]:		CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TA)
	220	0.358090	0.0	6.200000	1.000000	0.507	6.951	88.5	2.8617	8	30
	71	0.158760	0.0	10.810000	0.000000	0.413	5.961	17.5	5.2873	4	30!
	240	0.113290	30.0	4.930000	0.069959	0.428	6.897	54.3	6.3361	6	300
	6	0.088290	12.5	7.870000	0.069959	0.524	6.012	66.6	5.5605	5	31:
	417	25.940600	0.0	18.100000	0.000000	0.679	5.304	89.1	1.6475	24	660
	323	0.283920	0.0	7.380000	0.000000	0.493	5.708	74.3	4.7211	5	28
	192	3.611874	45.0	3.440000	0.000000	0.437	7.178	26.3	6.4798	5	398
	117	0.150980	0.0	10.010000	0.000000	0.547	6.021	82.6	2.7474	6	437
	47	0.229270	0.0	11.083992	0.000000	0.448	6.030	85.5	5.6894	3	233
	172	0.139140	0.0	4.050000	0.000000	0.510	5.572	88.5	2.5961	5	290

404 rows × 13 columns

```
In [11]: from sklearn.linear_model import LinearRegression
   model = LinearRegression()
   model.fit(x_train.values, y_train.values)
```

Out[11]: LinearRegression © CLinearRegression()

```
In [12]: y_train_pred = model.predict(x_train.values)
y_test_pred = model.predict(x_test.values)
```

In [13]: y_train_pred

```
Out[13]: array([32.66475357, 22.54760255, 27.92272071, 23.625425 , 6.45861412,
                 13.93671818, 21.99813652, 29.37255546, 32.26120163, 12.96973779,
                 19.94017093, 21.45691431, 12.8839238 , 23.92943611, 5.97521446,
                 18.93574189, 9.22524241, 45.33468083, 30.74460224, 17.25459323,
                 17.76350511, 21.90996473, 23.29496036, 18.97171185, 34.91668639,
                 13.50648784, 20.87214382, 35.52772443, 19.06816761, 13.54186364,
                 13.76659184, 22.07218103, 14.98269752, 31.34278627, 25.39744809,
                 16.1870452 , 24.86653661, 9.63533679, 15.04079584, 21.9231712 ,
                 33.08970263, 28.30793582, 26.6815381 , 15.31770869, 31.83213513,
                 25.30973944, 14.15787572, 7.92149041, 27.76577059, 25.30401239,
                  4.99697272, 28.22349867, 16.77086549, 29.86368234, 19.11821305,
                 16.15376997, 18.41077744, 12.7567695 , 8.79391708, 19.05618572,
                 35.71550638, 32.72210532, 23.71015905, 20.18648812, 22.93000843,
                 26.44974296, 21.0615435 , 17.69830015, 32.39856766, 10.49365749,
                 19.10820204, 31.36688112, 18.96836882, 15.76882773, 18.94519156,
                 15.15847982, 24.11855199, 23.38207934, 17.3295206 , 13.10475782,
                 20.54654327, 23.99183807, 17.46003754, 25.57826811, 22.75564999,
                 27.96955482, 36.79911276, 16.29320092, 11.96928193, 34.98157367,
                 31.2956174 , 20.32883487, 39.91898815, 28.63629469, 28.52460167,
                 17.68554714, 26.84504736, 40.39261223, 27.46393012, 17.137448
                 37.46198772, 35.70706734, 14.06849289, 27.49265535, 24.01488559,
                 24.968266 , 21.24194434, 23.22159002, 27.94396566, 29.45052215,
                 13.83197008, 26.09853134, 22.80277822, 15.43676479, 13.97762078,
                 25.50947859, 19.48687456, 30.44350731, 9.85025779, 24.27331225,
                 17.03316039, 16.85546978, 22.75870841, 21.67495995, 12.0451091,
                 25.08931078, 28.51617952, 22.65604474, 12.19565184, 24.93552503,
                 26.58866568, 25.7705037 , 23.51770572, 25.60406486, 19.33663113,
                 20.71038807, 36.06303681, 20.99110962, 36.30038206, 25.71105746,
                 20.83663618, 15.40620166, 32.04196757, 21.32944187, 28.06820878,
                 16.71628892, 32.72246926, 14.11501893, 1.48726866, 19.44446647,
                 13.70894591, 37.55581629, 16.17814528, 14.35177526, 26.62492959,
                 23.51386813, 17.58187046, 31.19908409, 25.40403936, 27.60077495,
                 24.60481019, 22.86284304, 22.41781158, 11.11394138, 20.73667755,
                 11.53702529, 17.62501807, 12.30439483, 27.55821134, 14.98300245,
                 15.86027234, 28.61576048, 14.2952606 , 21.67971867, 12.6258741 ,
                 16.4079762 , 23.3856092 , 21.14944384, 14.80325171, 17.41895461,
                 14.95913785, 26.00657216, 12.54531255, 35.34831489, 14.44057714,
                 43.19084197, 31.60225234, 34.73011197, 22.06549381, 15.60881839,
                 26.86711745, 29.08585412, 13.6020358 , 26.62936499, 36.01124274,
                 16.70502674, 11.53256998, 34.49979514, 36.02843502, 17.94009918,
                 21.23466028, 20.40463658, 24.44503261, 19.5133743 , 27.09874464,
                 -4.35147473, 20.83165704, 32.92028686, 35.38683579, 25.09427472,
                 26.70282143, 20.19959097, 21.47888434, 16.00569814, 17.82411768,
                 21.22008077, 27.93097558, 19.83756506, 6.96957231, 16.07007846,
                 32.24463684, 35.49232568, 16.35221777, 18.86660548, 22.33681004,
                 6.36701386, 21.42824658, 23.58756671, 15.85947528, 18.40347802,
                 23.21231647, 27.04107566, 25.85228541, 32.73542374, 14.68833813,
                 28.93776787, 24.93274252, 20.81833898, 38.5230573 , 21.98860617,
                 23.59465047, 22.5915317 , 12.04014312, 20.06564025, 33.38017755,
                 24.71481958, 17.76569636, 33.11025806, 22.08990019, 28.73314095,
                 32.06058555, 36.47410146, 21.66024396, 24.05946361, 23.1696524,
                 32.0392582 , 22.19777583 , 18.26850685 , 21.89849598 , 29.33625032 ,
                 22.89607891, 21.98997098, 16.9860062 , 17.32403378, 16.95639466,
                 16.85650119, 16.67797165, 31.93962811, 23.06540444, 17.48840171,
                 19.06901983, 34.20518995, 13.92961901, 26.13376776, 16.98878425,
                 30.66239417, 29.94233138, 23.11218925, 20.15856534, 36.1991215 ,
```

```
20.53732926, 33.46685682, 21.36666915, 31.50394589, 30.09643429,
                 37.30345689, 25.71187931, 20.88627074, 29.11779833, 15.97053177,
                 25.91460539, 21.48266931, 29.92258458, 10.29951201, 31.2335619,
                  6.23492083, 15.21271299, 20.37088814, 35.76335679, 31.79152748,
                 12.22466543, 13.67981558, 21.98115843, 34.69693923, 18.75887204,
                 18.38805663, 14.72038213, 25.298637 , 40.95303912, 25.39285876,
                 42.04574424, 25.45406537, 20.88512916, 11.87425463, 15.8524752,
                 14.04805806, 18.44010864, 3.0485567, 27.64307522, 26.43528674,
                 41.75366915, 21.65531543, 21.07349544, 34.06526064, 32.73156545,
                 9.46651233, 24.80881062, 43.77303166, 21.73707107, 17.66971023,
                 26.3438443 , 18.42979496 , 6.36551293 , 18.82126102 , 35.63631667 ,
                 16.10479431, 23.92928569, 13.10217538, 24.45122275, 18.1733537,
                 17.16243484, 18.27651925, 32.98016225, 19.24753909, 29.60002147,
                 31.76788565, 43.45274978, 18.39829662, 15.77094546, 38.32218921,
                 17.48890093, 10.4567689 , 14.58291612, 25.32215453, 19.39038299,
                 16.3931107 , 26.4981961 , 15.21641805, 5.99624195, 18.72341049,
                 11.00565791, 28.43940705, 4.77977595, 28.49842529, 32.7749635,
                 22.71279076, 16.39172307, 17.70378296, 21.20658832, 33.91580167,
                 28.3465387 , 19.40882851, 20.33459157, 6.88538961, 29.06990391,
                                                     , 24.48063603, 19.39198825,
                 25.17119662, 22.45813448, 13.53193
                 8.76491126, 26.75622464, 15.84398325, 31.60691698, 33.593711
                 24.99304656, 18.41963172, 30.42821065, 21.24240722, 26.09533173,
                 24.3152015 , 31.15075678, 24.44459048, 31.50609222, 17.48431929,
                 19.54265449, 18.76350478, 41.32674441, 25.65378283, 19.17369501,
                 33.43036639, 23.57405103, 18.03491836, 23.12408505])
In [14]: model.predict([[0.00632, 18.0, 2.31, 0.0, 0.538, 6.575, 65.2, 4.0900, 1.0, 2
Out[14]: array([30.66239417])
In [15]:
         y test pred
Out[15]: array([26.38244903, 22.48142477, 28.9150927, 11.51311019, 21.55596144,
                 19.39569936, 20.24684687, 21.42696075, 19.34116377, 19.68860665,
                  4.16870305, 15.92252628, 16.82679893, 5.33336826, 39.08414873,
                 33.13236137, 21.86276797, 36.51777073, 31.70654122, 23.61115389,
                 24.93638896, 23.41076277, 20.78702452, 30.60639121, 22.78779124,
                 8.34225949, 17.46512364, 17.83410415, 35.88821362, 21.01664412,
                 17.72326132, 17.42852689, 19.15036073, 23.30738941, 30.794513
                 19.22448935, 11.2795003 , 23.85638994, 17.65920118, 15.27836886,
                 26.42841366, 21.60671364, 23.86193607, 14.66954994, 23.89469709,
                 24.73879055, 19.95659189, 23.04414555, 10.49937272, 24.42914586,
                 23.54775421, 18.98321991, 24.46540719, 31.13192075, 12.69208308,
                 22.46866718, 21.36471519, 15.97301174, 12.04261875, 22.54049973,
                 18.20501757, 21.91741417, 32.62538174, 31.43041916, 17.56480551,
                 33.25033675, 21.22233526, 19.67388402, 19.97829012, 24.072029
                 22.59941828, 24.13111571, 30.83037261, 28.84034313, 25.01388616,
                 5.68203177, 38.85648038, 24.12681527, 27.6468537, 19.64454572,
                 28.67492372, 18.62889753, 17.47846418, 37.87598241, 39.37598
                 24.26550781, 25.2577793 , 16.66424155, 25.32136054, 16.69295872,
                 16.37096093, 13.29262768, 24.62082096, 31.00936186, 23.93931001,
                 20.41960539, 0.89355306, 25.32071273, 15.40869795, 17.51079031,
                 25.88556368, 22.24344716])
In [16]: y train
```

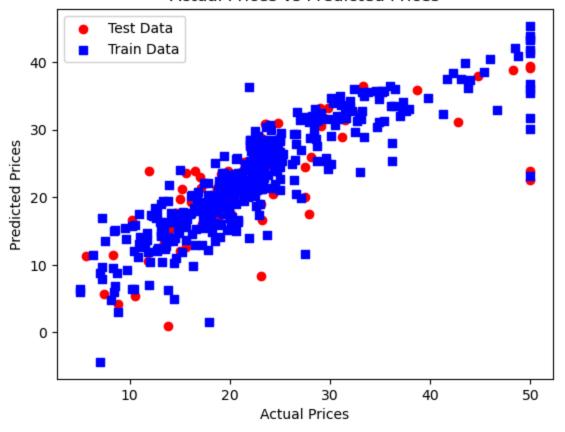
```
Out[16]: 220
                 26.7
         71
                 21.7
          240
                 22.0
          6
                 22.9
          417
                 10.4
                 . . .
          323
                 18.5
          192
                 36.4
          117
                 19.2
          47
                 16.6
          172
                 23.1
         Name: MEDV, Length: 404, dtype: float64
In [17]: y_test
Out[17]: 329
                 22.6
          371
                 50.0
          219
                 23.0
          403
                 8.3
          78
                 21.2
                 . . .
          56
                 24.7
          455
                 14.1
                 18.7
          60
          213
                 28.1
          108
                 19.8
         Name: MEDV, Length: 102, dtype: float64
In [18]: df1 = pd.DataFrame({'Actual': y_train, 'Predicted': y_train_pred})
         df2 = pd.DataFrame({'Actual': y_test, 'Predicted': y_test_pred})
In [19]: df1
Out[19]:
               Actual Predicted
         220
                 26.7 32.664754
                 21.7 22.547603
           71
         240
                 22.0 27.922721
                 22.9 23.625425
         417
                 10.4 6.458614
         323
                 18.5 19.173695
          192
                 36.4 33.430366
         117
                 19.2 23.574051
                 16.6 18.034918
           47
         172
                 23.1 23.124085
```

 $404 \text{ rows} \times 2 \text{ columns}$

```
Actual Predicted
Out[20]:
         329
                 22.6 26.382449
         371
                 50.0 22.481425
                 23.0 28.915093
         219
          403
                  8.3 11.513110
           78
                 21.2 21.555961
                 24.7 25.320713
           56
                 14.1 15.408698
          455
                 18.7 17.510790
           60
                 28.1 25.885564
         213
         108
                 19.8 22.243447
         102 \text{ rows} \times 2 \text{ columns}
In [21]: from sklearn.metrics import mean squared error, r2 score
         mse = mean squared error(y test, y test pred)
Out[21]: 35.89364343464623
In [22]: mse = mean squared error(y train, y train pred)
         mse
Out[22]: 19.805844906182962
In [23]: r2 = r2_score(y_test, y_test_pred)
         r2
Out[23]: 0.5592001535561668
In [24]: plt.scatter(y_test, y_test_pred, c='red', marker='o', label='Test Data')
         plt.scatter(y train, y train pred, c='blue', marker='s', label='Train Data')
         plt.xlabel('Actual Prices')
         plt.ylabel('Predicted Prices')
         plt.title('Actual Prices vs Predicted Prices')
         plt.legend(loc='upper left')
         plt.plot()
         plt.show()
```

In [20]: df2

Actual Prices vs Predicted Prices



In []:

This notebook was converted with convert.ploomber.io