

```
In [1]: 1 import pandas as pd
        2 import numpy as np
```

```
In [2]: 1 df=pd.read_csv(r"C:\Users\kaush\Downloads\housing.csv")
```

```
In [3]: 1 df
```

Out[3]:

	RM	LSTAT	PTRATIO	MEDV
0	6.575	4.98	15.3	504000.0
1	6.421	9.14	17.8	453600.0
2	7.185	4.03	17.8	728700.0
3	6.998	2.94	18.7	701400.0
4	7.147	5.33	18.7	760200.0
...
484	6.593	9.67	21.0	470400.0
485	6.120	9.08	21.0	432600.0
486	6.976	5.64	21.0	501900.0
487	6.794	6.48	21.0	462000.0
488	6.030	7.88	21.0	249900.0

489 rows × 4 columns

```
In [4]: 1 df.mean()
```

Out[4]: RM 6.240288
LSTAT 12.939632
PTRATIO 18.516564
MEDV 454342.944785
dtype: float64

```
In [5]: 1 df.loc[:, 'RM'].mean()
```

Out[5]: 6.240288343558283

```
In [6]: 1 df.mean(axis=1)[0:4]
```

Out[6]: 0 126006.71375
1 113408.34025
2 182182.25375
3 175357.15950
dtype: float64

```
In [7]: 1 df.median()
```

```
Out[7]: RM                6.185  
LSTAT            11.690  
PTRATIO          19.100  
MEDV        438900.000  
dtype: float64
```

```
In [8]: 1 df.loc[:, 'RM'].median()
```

```
Out[8]: 6.185
```

```
In [9]: 1 df.median(axis=1)[0:4]
```

```
Out[9]: 0    10.9375  
1    13.4700  
2    12.4925  
3    12.8490  
dtype: float64
```

```
In [10]: 1 df.mode()
```

```
Out[10]:
```

	RM	LSTAT	PTRATIO	MEDV
0	5.713	6.36	20.2	525000.0
1	6.127	7.79	NaN	NaN
2	6.167	8.05	NaN	NaN
3	6.229	14.10	NaN	NaN
4	6.405	18.13	NaN	NaN
5	6.417	NaN	NaN	NaN

```
In [11]: 1 df.loc[:, 'RM'].mode()
```

```
Out[11]: 0    5.713  
1    6.127  
2    6.167  
3    6.229  
4    6.405  
5    6.417  
Name: RM, dtype: float64
```

```
In [12]: 1 df.min()
```

```
Out[12]: RM                3.561  
LSTAT            1.980  
PTRATIO          12.600  
MEDV        105000.000  
dtype: float64
```

```
In [13]: 1 df.loc[:, 'RM'].min(skipna=False)
```

```
Out[13]: 3.561
```

```
In [14]: 1 df.max()
```

```
Out[14]: RM                8.398
LSTAT          37.970
PTRATIO        22.000
MEDV         1024800.000
dtype: float64
```

```
In [16]: 1 df.loc[:, 'RM'].max(skipna=False)
```

```
Out[16]: 8.398
```

```
In [17]: 1 df.std()
```

```
Out[17]: RM                0.643650
LSTAT          7.081990
PTRATIO        2.111268
MEDV         165340.277653
dtype: float64
```

```
In [18]: 1 df.loc[:, 'RM'].std()
```

```
Out[18]: 0.6436497627572431
```

```
In [19]: 1 df.std(axis=1)[0:4]
```

```
Out[19]: 0    251995.524207
1    226794.439885
2    364345.164214
3    350695.227064
dtype: float64
```

```
In [20]: 1 df.groupby(['LSTAT'])['RM'].mean()
```

```
Out[20]: LSTAT
1.98      7.024
2.47      8.337
2.87      7.178
2.94      6.998
2.98      6.854
...
34.37     4.628
34.41     5.019
34.77     4.906
36.98     4.519
37.97     4.138
Name: RM, Length: 442, dtype: float64
```

```
In [22]: 1 from scipy import stats
2 df=pd.read_csv(r"C:\Users\kaush\Downloads\housing.csv")
3 z= np.abs(stats.zscore(df['RM']))
4 print(z)
```

```
0      0.520554
1      0.281048
2      1.469245
3      1.178417
4      1.410146
...
484    0.548548
485    0.187076
486    1.144202
487    0.861150
488    0.327047
Name: RM, Length: 489, dtype: float64
```

```
In [23]: 1 threshold=0.18
2 sample_outliers=np.where(z<threshold)
3 sample_outliers
```

```
Out[23]: (array([ 7, 22, 42, 43, 58, 65, 73, 74, 75, 76, 77, 78, 82,
83, 93, 94, 96, 103, 104, 107, 109, 110, 114, 116, 131, 135,
139, 140, 145, 154, 155, 160, 164, 173, 177, 181, 195, 199, 208,
214, 239, 261, 275, 276, 277, 282, 287, 302, 315, 318, 319, 322,
323, 345, 347, 350, 362, 376, 384, 410, 411, 413, 417, 420, 425,
429, 430, 431, 435, 441, 445, 447, 453, 454, 458, 461, 462, 463,
468], dtype=int64),)
```

```
In [24]: 1 sorted_rscore= sorted(df['RM'])
2 sorted_rscore
3 q1 =np.percentile(sorted_rscore,6)
4 q3 =np.percentile(sorted_rscore,9)
5 print(q1,q3)
```

```
5.3925199999999999 5.5580799999999995
```

```
In [25]: 1 IQR = q3-q1
2 lwr_bound =q1-(1.5*IQR)
3 upr_bound =q3+(1.5*IQR)
4 print(lwr_bound, upr_bound)
```

```
5.1441799999999999 5.8064199999999999
```

```
In [26]: 1 r_outliers =[]
2 for i in sorted_rscore:
3     if(i<lwr_bound or i>upr_bound):
4         r_outliers.append(i)
5 print(r_outliers)
```

```
[3.561, 3.863, 4.138, 4.138, 4.368, 4.519, 4.628, 4.652, 4.88, 4.903, 4.906,
4.926, 4.963, 4.973, 5.0, 5.012, 5.019, 5.036, 5.093, 5.807, 5.813, 5.813, 5.
818, 5.822, 5.834, 5.836, 5.837, 5.841, 5.85, 5.851, 5.852, 5.854, 5.854, 5.8
56, 5.856, 5.857, 5.859, 5.868, 5.869, 5.87, 5.871, 5.872, 5.874, 5.875, 5.87
6, 5.877, 5.878, 5.879, 5.88, 5.884, 5.885, 5.887, 5.888, 5.888, 5.889, 5.89
1, 5.895, 5.896, 5.898, 5.905, 5.913, 5.914, 5.92, 5.924, 5.926, 5.926, 5.92
7, 5.928, 5.933, 5.935, 5.935, 5.936, 5.936, 5.942, 5.949, 5.95, 5.951, 5.95
2, 5.957, 5.96, 5.961, 5.961, 5.963, 5.965, 5.966, 5.966, 5.968, 5.972, 5.97
6, 5.981, 5.983, 5.983, 5.985, 5.986, 5.987, 5.99, 5.998, 6.003, 6.004, 6.00
4, 6.006, 6.009, 6.009, 6.012, 6.014, 6.015, 6.019, 6.02, 6.021, 6.023, 6.02
7, 6.03, 6.03, 6.031, 6.037, 6.041, 6.047, 6.051, 6.059, 6.064, 6.065, 6.066,
6.069, 6.072, 6.081, 6.083, 6.086, 6.092, 6.095, 6.096, 6.096, 6.101, 6.103,
6.108, 6.108, 6.112, 6.113, 6.114, 6.115, 6.12, 6.121, 6.122, 6.122, 6.127,
6.127, 6.127, 6.129, 6.13, 6.137, 6.14, 6.142, 6.144, 6.144, 6.145, 6.151, 6.
152, 6.152, 6.153, 6.162, 6.162, 6.163, 6.164, 6.167, 6.167, 6.167, 6.169, 6.
172, 6.174, 6.176, 6.182, 6.185, 6.185, 6.193, 6.193, 6.195, 6.202, 6.208, 6.
209, 6.209, 6.211, 6.211, 6.212, 6.219, 6.223, 6.226, 6.229, 6.229, 6.229, 6.
23, 6.232, 6.24, 6.242, 6.245, 6.249, 6.25, 6.251, 6.251, 6.254, 6.266, 6.27
3, 6.279, 6.286, 6.29, 6.297, 6.301, 6.302, 6.31, 6.312, 6.312, 6.315, 6.315,
6.316, 6.317, 6.319, 6.326, 6.326, 6.333, 6.335, 6.341, 6.343, 6.345, 6.348,
6.358, 6.372, 6.373, 6.375, 6.376, 6.376, 6.377, 6.38, 6.38, 6.382, 6.383, 6.
389, 6.393, 6.395, 6.398, 6.402, 6.404, 6.405, 6.405, 6.405, 6.406, 6.411, 6.
415, 6.416, 6.417, 6.417, 6.417, 6.421, 6.425, 6.426, 6.43, 6.431, 6.431, 6.4
33, 6.434, 6.436, 6.437, 6.438, 6.442, 6.453, 6.454, 6.456, 6.458, 6.459, 6.4
61, 6.471, 6.474, 6.481, 6.482, 6.484, 6.485, 6.487, 6.49, 6.495, 6.495, 6.5
1, 6.511, 6.513, 6.516, 6.525, 6.538, 6.54, 6.545, 6.546, 6.549, 6.552, 6.55
6, 6.563, 6.565, 6.567, 6.575, 6.579, 6.59, 6.593, 6.595, 6.604, 6.606, 6.61
6, 6.618, 6.619, 6.625, 6.629, 6.63, 6.63, 6.631, 6.635, 6.635, 6.642, 6.649,
6.655, 6.657, 6.674, 6.678, 6.696, 6.701, 6.715, 6.718, 6.726, 6.727, 6.727,
6.728, 6.728, 6.739, 6.749, 6.75, 6.758, 6.762, 6.77, 6.781, 6.782, 6.782, 6.
794, 6.794, 6.8, 6.812, 6.816, 6.824, 6.826, 6.833, 6.842, 6.849, 6.852, 6.85
4, 6.86, 6.861, 6.871, 6.874, 6.879, 6.897, 6.939, 6.943, 6.951, 6.951, 6.95
7, 6.968, 6.968, 6.975, 6.976, 6.98, 6.98, 6.982, 6.998, 7.007, 7.014, 7.024,
7.041, 7.061, 7.079, 7.088, 7.104, 7.107, 7.135, 7.147, 7.148, 7.155, 7.163,
7.178, 7.185, 7.185, 7.203, 7.206, 7.236, 7.241, 7.249, 7.267, 7.274, 7.287,
7.313, 7.327, 7.333, 7.358, 7.393, 7.412, 7.416, 7.42, 7.454, 7.47, 7.52, 7.6
1, 7.645, 7.686, 7.691, 7.765, 7.82, 7.82, 7.853, 8.04, 8.069, 8.247, 8.259,
8.266, 8.337, 8.398]
```