Bengaluru House Price Predictor

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1 Bengaluru House Price Prediction Model By Navjoth Singh

```
[1]: import pandas as pd
     import numpy as np
[2]: data = pd.read_csv("C:/Users/hp/OneDrive/Desktop/Data Science/Data for Practice/
      →Bengaluru_House_Data.csv")
[3]: data.head()
[3]:
                                availability
                                                               location
                                                                               size
                   area_type
        Super built-up
                        Area
                                      19-Dec
                                              Electronic City Phase II
                                                                              2 BHK
     1
                  Plot
                        Area
                               Ready To Move
                                                       Chikka Tirupathi
                                                                         4 Bedroom
     2
              Built-up
                        Area
                               Ready To Move
                                                            Uttarahalli
                                                                              3 BHK
     3 Super built-up
                               Ready To Move
                                                     Lingadheeranahalli
                        Area
                                                                              3 BHK
        Super built-up
                        Area
                               Ready To Move
                                                               Kothanur
                                                                              2 BHK
        society total_sqft
                             bath
                                   balcony
                                             price
     0
        Coomee
                      1056
                              2.0
                                       1.0
                                             39.07
        Theanmp
                      2600
                              5.0
                                       3.0
                                            120.00
     1
                      1440
                                             62.00
     2
            NaN
                              2.0
                                       3.0
     3
       Soiewre
                      1521
                              3.0
                                       1.0
                                             95.00
            NaN
                      1200
                              2.0
                                       1.0
                                             51.00
[4]: data.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 13320 entries, 0 to 13319
    Data columns (total 9 columns):
     #
         Column
                        Non-Null Count
                                        Dtype
         _____
                        _____
     0
                        13320 non-null
                                        object
         area_type
                        13320 non-null
     1
         availability
                                        object
     2
         location
                        13319 non-null
                                        object
     3
         size
                        13304 non-null
                                        object
     4
         society
                        7818 non-null
                                        object
     5
         total_sqft
                        13320 non-null
                                        object
                        13247 non-null
         bath
                                        float64
```

```
balcony
                       12711 non-null float64
                       13320 non-null float64
         price
    dtypes: float64(3), object(6)
    memory usage: 936.7+ KB
[5]: data.shape
[5]: (13320, 9)
[6]: data.columns
[6]: Index(['area_type', 'availability', 'location', 'size', 'society',
            'total_sqft', 'bath', 'balcony', 'price'],
           dtype='object')
[7]: data.area_type
[7]: 0
              Super built-up
                              Area
     1
                        Plot
                             Area
     2
                    Built-up Area
     3
              Super built-up Area
     4
              Super built-up Area
     13315
                    Built-up Area
     13316
              Super built-up Area
     13317
                    Built-up Area
     13318
              Super built-up Area
              Super built-up Area
     13319
     Name: area_type, Length: 13320, dtype: object
[8]: data.area_type.value_counts()
[8]: area_type
     Super built-up Area
                             8790
     Built-up Area
                             2418
     Plot Area
                             2025
     Carpet Area
                               87
     Name: count, dtype: int64
[9]: data["area_type"].value_counts()
[9]: area_type
     Super built-up
                   Area
                             8790
     Built-up Area
                             2418
    Plot Area
                             2025
     Carpet Area
                               87
    Name: count, dtype: int64
```

```
[10]: for column in data.columns:
         print(data[column].value_counts())
         print("*"*20)
     area_type
     Super built-up Area
                            8790
     Built-up Area
                            2418
     Plot Area
                            2025
     Carpet Area
                              87
     Name: count, dtype: int64
     *******
     availability
     Ready To Move
                     10581
     18-Dec
                        307
     18-May
                        295
     18-Apr
                        271
                        200
     18-Aug
                          1
     15-Aug
     17-Jan
                         1
     16-Nov
                          1
     16-Jan
                         1
     14-Jul
                          1
     Name: count, Length: 81, dtype: int64
     *******
     location
                                      540
     Whitefield
     Sarjapur Road
                                      399
     Electronic City
                                      302
     Kanakpura Road
                                      273
     Thanisandra
                                      234
     Bapuji Layout
                                        1
     1st Stage Radha Krishna Layout
                                        1
     BEML Layout 5th stage
                                        1
     singapura paradise
                                        1
     Abshot Layout
     Name: count, Length: 1305, dtype: int64
     ******
     size
     2 BHK
                  5199
     3 ВНК
                   4310
     4 Bedroom
                    826
     4 BHK
                    591
     3 Bedroom
                    547
     1 BHK
                    538
                    329
     2 Bedroom
```

297

5 Bedroom

```
6 Bedroom
               191
1 Bedroom
               105
8 Bedroom
                84
7 Bedroom
                83
5 BHK
                59
9 Bedroom
                46
6 BHK
                30
7 BHK
                17
1 RK
                13
10 Bedroom
                12
9 BHK
                 8
8 BHK
                 5
                 2
11 BHK
                 2
11 Bedroom
10 BHK
                 2
14 BHK
                 1
13 BHK
                 1
12 Bedroom
                 1
27 BHK
                 1
43 Bedroom
                 1
16 BHK
                 1
19 BHK
18 Bedroom
                 1
Name: count, dtype: int64
*******
society
GrrvaGr
           80
           76
PrarePa
Sryalan
           59
Prtates
           59
GMown E
           56
           . .
Amionce
            1
JaghtDe
            1
Jauraht
            1
Brity U
            1
RSntsAp
            1
Name: count, Length: 2688, dtype: int64
************
total_sqft
1200
        843
1100
        221
1500
        205
2400
        196
600
        180
3580
          1
2461
          1
```

```
1437
         1
2155
         1
4689
         1
Name: count, Length: 2117, dtype: int64
******
bath
2.0
       6908
3.0
       3286
4.0
       1226
1.0
        788
5.0
        524
6.0
        273
7.0
        102
8.0
         64
9.0
         43
10.0
         13
12.0
          7
13.0
          3
11.0
          3
          2
16.0
27.0
          1
40.0
          1
15.0
          1
14.0
          1
18.0
          1
Name: count, dtype: int64
******
balcony
2.0
      5113
1.0
      4897
3.0
      1672
      1029
0.0
Name: count, dtype: int64
******
price
75.00
         310
65.00
         302
55.00
         275
60.00
         270
45.00
         240
351.00
           1
54.10
           1
80.64
           1
32.73
           1
           1
Name: count, Length: 1994, dtype: int64
*******
```

```
[11]: data.isna().sum()
[11]: area_type
                       0
     availability
                       0
     location
                       1
     size
                      16
                    5502
     society
     total_sqft
                       0
     bath
                      73
     balcony
                     609
     price
                       0
     dtype: int64
        Data Cleaning
[12]: data.drop(columns_
      # Dropping columns that is no longer required
[13]: data.describe()
[13]:
                   bath
                                price
           13247.000000
                         13320.000000
     count
               2.692610
                           112.565627
     mean
     std
                1.341458
                           148.971674
     min
                1.000000
                             8.000000
     25%
               2.000000
                            50.000000
     50%
               2.000000
                            72.000000
     75%
                3.000000
                           120.000000
               40.000000
                          3600.000000
     max
[14]: data.isnull().sum()
[14]: location
     size
                  16
     total_sqft
                   0
     bath
                  73
     price
                   0
     dtype: int64
[15]: data['location'].value_counts().head()
[15]: location
     Whitefield
                       540
                       399
     Sarjapur Road
     Electronic City
                       302
     Kanakpura Road
                       273
```

```
234
      Thanisandra
      Name: count, dtype: int64
[16]: data['location'].fillna('Whitefield',inplace=True) # 1 is added to Whitefield
[17]: data['location'].value_counts().head()
[17]: location
      Whitefield
                         541
                         399
      Sarjapur Road
      Electronic City
                         302
      Kanakpura Road
                         273
      Thanisandra
                         234
      Name: count, dtype: int64
[18]: data['bath'].value_counts().head(5)
[18]: bath
      2.0
             6908
      3.0
             3286
      4.0
             1226
      1.0
              788
      5.0
              524
      Name: count, dtype: int64
[19]: data['bath'].fillna(data['bath'].mode()[0],inplace=True)
[20]: | data['bath'].value_counts().head()
[20]: bath
      2.0
             6981
      3.0
             3286
      4.0
             1226
      1.0
              788
      5.0
              524
      Name: count, dtype: int64
[21]: data['size'].value_counts().head()
[21]: size
      2 BHK
                   5199
      3 BHK
                   4310
      4 Bedroom
                    826
      4 BHK
                    591
      3 Bedroom
                    547
      Name: count, dtype: int64
[22]: data['size'].fillna(data['size'].mode()[0],inplace=True)
```

```
[23]: data['size'].value_counts().head()
[23]: size
     2 BHK
                  5215
      3 ВНК
                   4310
      4 Bedroom
                    826
      4 BHK
                    591
                    547
      3 Bedroom
      Name: count, dtype: int64
[24]: data['bhk']=data['size'].str.split().str.get(0).astype(int) # New column_
       ⇔created with name BHK
[25]: data.head()
[25]:
                         location
                                        size total_sqft bath
                                                                price
                                                                       bhk
      O Electronic City Phase II
                                       2 BHK
                                                   1056
                                                          2.0
                                                                39.07
                                                                         2
                 Chikka Tirupathi 4 Bedroom
                                                   2600
                                                          5.0 120.00
      1
      2
                      Uttarahalli
                                       3 BHK
                                                   1440
                                                          2.0
                                                                62.00
                                                                         3
      3
               Lingadheeranahalli
                                       3 BHK
                                                   1521
                                                          3.0
                                                                95.00
                                                                         3
      4
                         Kothanur
                                       2 BHK
                                                   1200
                                                          2.0
                                                                51.00
                                                                         2
[26]: data[data.bhk > 20]
                             # We have to fix them because more than 20 Room bedroom_
       ⇔is hypothetical - 1
[26]:
                             location
                                             size total_sqft bath price bhk
      1718 2Electronic City Phase II
                                           27 BHK
                                                        8000
                                                             27.0 230.0
                                                                            27
      4684
                          Munnekollal 43 Bedroom
                                                        2400 40.0 660.0
[27]: data['total_sqft'].unique()
                                      # We have to fix them because hyphen are there__
       →- 2
[27]: array(['1056', '2600', '1440', ..., '1133 - 1384', '774', '4689'],
            dtype=object)
[28]: def convertRange(x):
                               # Fixing - 2
          temp = x.split('-')
          if len(temp) == 2:
              return (float(temp[0])+float(temp[1]))/2
          try:
             return float(x)
          except:
              return None
[29]: data['total_sqft']=data['total_sqft'].apply(convertRange)
```

```
[30]: data['total_sqft'].unique() # You may see that '1133 - 1384' changes to 1258.
       5
[30]: array([1056., 2600., 1440., ..., 1258.5, 774., 4689.])
[31]: # Price Per Square feet
      data['price_per_sqft']=data['price']*100000/data['total_sqft']
[32]: data.head()
[32]:
                          location
                                         size
                                                total_sqft
                                                            bath
                                                                    price
                                                                           bhk
         Electronic City Phase II
                                         2 BHK
                                                             2.0
                                                                    39.07
      0
                                                    1056.0
                                                                             2
      1
                 Chikka Tirupathi 4 Bedroom
                                                    2600.0
                                                             5.0
                                                                   120.00
                                                                             4
      2
                      Uttarahalli
                                         3 ВНК
                                                                    62.00
                                                                             3
                                                    1440.0
                                                             2.0
                                                                    95.00
      3
               Lingadheeranahalli
                                        3 BHK
                                                    1521.0
                                                              3.0
                                                                             3
      4
                          Kothanur
                                        2 BHK
                                                    1200.0
                                                             2.0
                                                                    51.00
                                                                             2
         price_per_sqft
      0
            3699.810606
      1
            4615.384615
      2
            4305.55556
      3
            6245.890861
      4
            4250.000000
[33]: data.describe()
[33]:
               total_sqft
                                    bath
                                                  price
                                                                   bhk
                                                                        price_per_sqft
      count
             13274.000000
                            13320.000000
                                          13320.000000
                                                         13320.000000
                                                                          1.327400e+04
                                             112.565627
      mean
              1559.626694
                                2.688814
                                                              2.802778
                                                                          7.907501e+03
      std
              1238.405258
                                1.338754
                                             148.971674
                                                              1.294496
                                                                          1.064296e+05
      min
                 1.000000
                                1.000000
                                               8.000000
                                                              1.000000
                                                                          2.678298e+02
      25%
              1100.000000
                                2.000000
                                              50.000000
                                                              2.000000
                                                                          4.266865e+03
      50%
              1276.000000
                                2.000000
                                              72.000000
                                                             3.000000
                                                                          5.434306e+03
      75%
                                3.000000
                                                                          7.311746e+03
              1680.000000
                                             120.000000
                                                              3.000000
             52272.000000
                               40.000000
                                            3600.000000
                                                             43.000000
                                                                          1.200000e+07
      max
[34]: data['location'].value counts()
      # Now here we cannot use all these location on the model so we have to reduce_
       \hookrightarrow this
      # what will do, which ever location is less than 10 will replace it with
       →'other'
[34]: location
      Whitefield
                                         541
                                         399
      Sarjapur Road
      Electronic City
                                         302
      Kanakpura Road
                                         273
```

```
Thanisandra
                                         234
      Bapuji Layout
                                           1
      1st Stage Radha Krishna Layout
                                           1
      BEML Layout 5th stage
                                           1
      singapura paradise
                                           1
      Abshot Layout
                                           1
      Name: count, Length: 1305, dtype: int64
[35]: data['location'] = data['location'].apply(lambda x : x.strip())
      # it uses a lambda function, which strips whitespace characters (like spaces, u
       \hookrightarrow tabs, or newlines)
      # from both ends of each string in the 'location' column.
[36]: data['location'].value_counts() # Now we can see that length reduce to 1294
       ⇔from 1305
[36]: location
      Whitefield
                                         542
      Sarjapur Road
                                         399
      Electronic City
                                         304
      Kanakpura Road
                                         273
      Thanisandra
                                         237
      Bapuji Layout
                                           1
      1st Stage Radha Krishna Layout
                                           1
      BEML Layout 5th stage
                                           1
      singapura paradise
                                           1
      Abshot Layout
      Name: count, Length: 1294, dtype: int64
[37]: location_counts = data['location'].value_counts()
      location_count_less_10 = location_counts[location_counts <= 10]</pre>
      location_count_less_10
[37]: location
      Dairy Circle
                                         10
      Nagappa Reddy Layout
                                         10
      Basapura
                                         10
      1st Block Koramangala
                                         10
      Sector 1 HSR Layout
                                         10
                                         . .
      Bapuji Layout
                                          1
      1st Stage Radha Krishna Layout
                                          1
      BEML Layout 5th stage
                                          1
      singapura paradise
                                          1
```

```
Abshot Layout
      Name: count, Length: 1053, dtype: int64
[38]: data['location'] = data['location'].apply(lambda x: 'other' if x in_
       →location count less 10 else x)
[39]: data['location'].value_counts()
[39]: location
      other
                            2885
      Whitefield
                             542
      Sarjapur Road
                             399
      Electronic City
                             304
      Kanakpura Road
                              273
      Nehru Nagar
                               11
      Banjara Layout
                               11
      LB Shastri Nagar
                               11
      Pattandur Agrahara
                               11
      Narayanapura
                               11
      Name: count, Length: 242, dtype: int64
```

3 Dealing with outliers

```
[40]: data.describe()
[40]:
               total_sqft
                                    bath
                                                  price
                                                                   bhk
                                                                        price_per_sqft
                            13320.000000
             13274.000000
                                           13320.000000
                                                          13320.000000
                                                                           1.327400e+04
      count
      mean
              1559.626694
                                                                          7.907501e+03
                                2.688814
                                             112.565627
                                                              2.802778
      std
              1238.405258
                                1.338754
                                             148.971674
                                                              1.294496
                                                                          1.064296e+05
      min
                 1.000000
                                1.000000
                                               8.000000
                                                              1.000000
                                                                          2.678298e+02
      25%
              1100.000000
                                2.000000
                                              50.000000
                                                              2.000000
                                                                          4.266865e+03
      50%
              1276.000000
                                2.000000
                                              72.000000
                                                              3.000000
                                                                          5.434306e+03
      75%
                                                                          7.311746e+03
              1680.000000
                                3.000000
                                             120.000000
                                                              3.000000
             52272.000000
                               40.000000
                                            3600.000000
                                                             43.000000
                                                                          1.200000e+07
      max
[41]: (data['total_sqft']/data['bhk']).describe()
                                                        # representing the average_
       ⇔square footage per room
[41]: count
               13274.000000
                 575.074878
      mean
      std
                 388.205175
      min
                    0.250000
      25%
                 473.333333
      50%
                 552.500000
      75%
                 625.000000
      max
               26136.000000
```

```
dtype: float64
[42]: data = data[((data['total sqft']/data['bhk']) >= 300)]
      # I selected only those room which are more than 300 sqrft
                   # earlier it was (13320,9)
[43]: data.shape
[43]: (12530, 7)
[44]: data.price_per_sqft.describe()
      # Now here is 176470.588235 a outlier because this is unexpected price for a_{\sqcup}
       \hookrightarrow sqft
[44]: count
                12530.000000
      mean
                 6303.979357
      std
                 4162.237981
      min
                 267.829813
      25%
                 4210.526316
      50%
                 5294.117647
      75%
                 6916.666667
               176470.588235
      max
      Name: price_per_sqft, dtype: float64
[45]: def remove outliers sqft(df):
          df_output = pd.DataFrame()
          for key, subdf in df.groupby('location'):
              m = np.mean(subdf.price_per_sqft)
              st = np.std(subdf.price_per_sqft)
              gen_df = subdf[(subdf.price_per_sqft > (m-st)) & (subdf.price_per_sqft_
              df_output = pd.concat([df_output, gen_df], ignore_index = True)
          return df output
      data = remove_outliers_sqft(data)
      \mbox{\it\# Creates an empty DataFrame $$ `df_output$$ `to store the filtered data.}
      # During each iteration of the loop, key will take the value of the 'location'
       ⇔('Whitefield', 'Sarjapur Road', 'Electronic City')
      # and subdf will contain the data(price_per_sqft) of the DataFrame_{\sqcup}
       ⇔corresponding to that 'location'
      # Grouping by 'location': Iterates through each group formed by grouping the
       →DataFrame `df` by the 'location' column using `df.groupby('location')`.
      # Outlier Removal: For each group:
```

```
# Calculates the mean (`m`) and standard deviation (`st`) of the
       → 'price_per_sqft' within that location.
         # Generates a new DataFrame `gen_df` containing rows where the
       →'price per sqft' falls within one standard deviation from the mean for that,
       \rightarrow location.
      # Appends the filtered data (`gen_df`) to the empty `df_output` DataFrame using \Box
       → `pd.concat() `.
[46]: data.describe()
                        # See price per sqft is changed now \mathscr E the rows is reduced to \Box
       →10301 from 12530
[46]:
               total_sqft
                                   bath
                                                                 bhk price_per_sqft
                                                 price
             10301.000000
                           10301.000000 10301.000000 10301.000000
                                                                         10301.000000
      count
              1508.440608
                               2.471702
                                             91.286372
                                                            2.574896
                                                                          5659.062876
      mean
      std
               880.694214
                               0.979449
                                             86.342786
                                                            0.897649
                                                                          2265.774749
                               1.000000
                                             10.000000
                                                            1.000000
     min
               300.000000
                                                                          1250.000000
      25%
              1110.000000
                               2.000000
                                             49.000000
                                                            2.000000
                                                                          4244.897959
      50%
              1286.000000
                               2.000000
                                             67.000000
                                                            2.000000
                                                                          5175.600739
      75%
              1650.000000
                               3.000000
                                            100.000000
                                                            3.000000
                                                                          6428.571429
     max
             30400.000000
                              16.000000
                                           2200.000000
                                                           16.000000
                                                                        24509.803922
[47]: def bhk_outlier_remover(df):
          exclude_indices = np.array([])
          for key, subdf in df.groupby('location'):
              bhk_stats = {}
              for bhk, bhkdf in subdf.groupby('bhk'):
                  bhk_stats[bhk] = { 'mean' : np.mean(bhkdf.price_per_sqft),
                                     'std' : np.std(bhkdf.price_per_sqft),
                                     'count' : bhkdf.shape[0] }
              print(key, bhk_stats)
[48]: test data = bhk outlier remover(data)
     1st Block Jayanagar {2: {'mean': 11983.805668016194, 'std': 0.0, 'count': 1}, 3:
     {'mean': 11756.16905248807, 'std': 701.6243657657865, 'count': 3}, 4: {'mean':
     15018.711280365416, 'std': 1.2278182423353805, 'count': 3}}
     1st Phase JP Nagar {1: {'mean': 5952.380952380952, 'std': 0.0, 'count': 1}, 2:
     {'mean': 7931.806799837383, 'std': 1534.1422783514056, 'count': 8}, 3: {'mean':
     9151.192151725822, 'std': 1054.731726021645, 'count': 7}, 4: {'mean':
     7537.92218148637, 'std': 1607.0591069513537, 'count': 3}, 5: {'mean':
     5666.66666666667, 'std': 0.0, 'count': 1}}
     2nd Phase Judicial Layout {2: {'mean': 3851.8518518517, 'std':
     497.593660834978, 'count': 3}, 3: {'mean': 3620.93991671624, 'std':
     241.87983343248052, 'count': 5}}
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Akshaya Nagar {2: {'mean': 4781.22783734999, 'std': 453.10974836521865, 'count':
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Banashankari {1: {'mean': 8333.3333333334, 'std': 0.0, 'count': 2}, 2:
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Banashankari Stage II {2: {'mean': 8701.636799298034, 'std': 1030.287972433991,
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Bharathi Nagar {2: {'mean': 4813.740209190311, 'std': 471.43026983759114,
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Bommasandra {2: {'mean': 3813.947572153106, 'std': 244.236658864445, 'count':
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'std': 0.0, 'count': 1}}
Varthur {2: {'mean': 4331.910572925388, 'std': 847.3795043583863, 'count': 32},
3: {'mean': 4836.478103600665, 'std': 892.2656889001146, 'count': 14}}
Varthur Road {2: {'mean': 3754.9654055540996, 'std': 648.3927903937736, 'count':
7}, 3: {'mean': 3800.411985861906, 'std': 850.7508048357686, 'count': 2}}
Vasanthapura {2: {'mean': 3558.48776757308, 'std': 145.0953916099316, 'count':
7}}
Vidyaranyapura {2: {'mean': 4639.399867778706, 'std': 870.5642061348254,
'count': 16}, 3: {'mean': 5418.695502028836, 'std': 956.0813080374586, 'count':
6}, 4: {'mean': 6562.5, 'std': 937.5, 'count': 2}}
Vijayanagar {1: {'mean': 7513.33715688642, 'std': 1470.1647292081748, 'count':
4}, 2: {'mean': 5936.626849748775, 'std': 879.0177188541079, 'count': 11}, 3:
{'mean': 6585.510613483119, 'std': 860.4381230168186, 'count': 17}, 4: {'mean':
12000.0, 'std': 0.0, 'count': 1}, 5: {'mean': 5208.333333333333, 'std': 0.0,
```

```
'count': 1}}
Vishveshwarya Layout {1: {'mean': 4000.0, 'std': 0.0, 'count': 1}, 4: {'mean':
4125.0, 'std': 375.0, 'count': 2}, 6: {'mean': 4380.952380952381, 'std': 0.0,
'count': 1}}
Vishwapriya Layout {2: {'mean': 3916.9969520750556, 'std': 177.84130871379733,
'count': 4}}
Vittasandra {2: {'mean': 5265.466588027838, 'std': 172.68575428865756, 'count':
31}, 3: {'mean': 5113.14174326139, 'std': 100.56430016629145, 'count': 7}}
Whitefield {1: {'mean': 5765.189888976462, 'std': 1080.9401060122182, 'count':
26, 2: {'mean': 4996.679555545075, 'std': 1261.8095382276895, 'count': 232}, 3:
{'mean': 5529.545307677018, 'std': 1325.8662058810887, 'count': 166}, 4:
{'mean': 6781.90623203223, 'std': 1157.6199849662542, 'count': 57}, 5: {'mean':
7836.874286815605, 'std': 794.7110211453303, 'count': 4}}
Yelachenahalli {2: {'mean': 5157.852477694467, 'std': 706.3466043126809,
'count': 7}, 3: {'mean': 5909.760492768157, 'std': 1348.1242403302067, 'count':
5}}
Yelahanka {1: {'mean': 5250.754432669086, 'std': 708.1864329017459, 'count':
16}, 2: {'mean': 4885.50009150766, 'std': 834.2738830938632, 'count': 78}, 3:
{'mean': 5121.284167853846, 'std': 819.858135054314, 'count': 65}, 4: {'mean':
5396.634458670808, 'std': 859.2313582889707, 'count': 9}, 5: {'mean': 5150.0,
'std': 350.0, 'count': 2}, 6: {'mean': 5555.5555555556, 'std': 0.0, 'count':
1}}
Yelahanka New Town {1: {'mean': 3711.0711616859157, 'std': 705.7257959858678,
'count': 12}, 2: {'mean': 5255.289619148657, 'std': 1329.2576472001515, 'count':
7}, 3: {'mean': 5244.8328585493755, 'std': 1308.1039897883682, 'count': 12}, 4:
{'mean': 5202.863961813842, 'std': 0.0, 'count': 1}}
Yelenahalli {2: {'mean': 3878.617021276596, 'std': 155.17896357207565, 'count':
7}, 3: {'mean': 3918.9457070707067, 'std': 143.76151982471873, 'count': 3}}
Yeshwanthpur {1: {'mean': 5660.648244572341, 'std': 397.81389558095225, 'count':
13}, 2: {'mean': 6313.582176150229, 'std': 1092.0986164191115, 'count': 23}, 3:
{'mean': 6136.843144871886, 'std': 999.1342919214298, 'count': 31}, 4: {'mean':
6388.8888888889, 'std': 277.777777777778, 'count': 2}, 6: {'mean': 7400.0,
'std': 0.0, 'count': 1}}
other {1: {'mean': 5321.358826419817, 'std': 2357.705900996028, 'count': 132},
2: {'mean': 5288.580186585407, 'std': 1981.8078111382472, 'count': 987}, 3:
{'mean': 6303.178181413831, 'std': 2552.0634251169135, 'count': 858}, 4:
{'mean': 8334.159251776744, 'std': 2892.9656869519777, 'count': 224}, 5:
{'mean': 7278.542458208993, 'std': 3393.9728786041032, 'count': 42}, 6: {'mean':
6975.5623689134, 'std': 3043.3045406755605, 'count': 34}, 7: {'mean':
4272.049254022647, 'std': 1900.0507646394972, 'count': 11}, 8: {'mean':
5384.347002204145, 'std': 1946.6298824910634, 'count': 7}, 9: {'mean':
5530.842391304348, 'std': 3234.230886145459, 'count': 4}, 10: {'mean':
9005.6818181818, 'std': 4630.681818181818, 'count': 2}, 11: {'mean': 4850.0,
'std': 2350.0, 'count': 2}, 13: {'mean': 5069.124423963133, 'std': 0.0, 'count':
1}, 16: {'mean': 5500.0, 'std': 0.0, 'count': 1}}
```

```
[49]: def bhk_outlier_remover(df):
          exclude_indices = np.array([])
          for key, subdf in df.groupby('location'):
              bhk_stats = {}
              for bhk, bhkdf in subdf.groupby('bhk'):
                  bhk_stats[bhk] = { 'mean' : np.mean(bhkdf.price_per_sqft),
                                    'std' : np.std(bhkdf.price per sqft),
                                    'count' : bhkdf.shape[0] }
              for bhk, bhkdf in subdf.groupby('bhk'):
                  stats = bhk_stats.get(bhk-1)
                  if stats and stats['count'] > 5:
                      exclude_indices = np.append(exclude_indices, bhkdf[bhkdf.
       →price_per_sqft < (stats['mean'])].index.values)</pre>
          return df.drop(exclude_indices, axis = 'index')
[50]: data = bhk_outlier_remover(data)
[51]: data.shape
[51]: (7361, 7)
[52]: data.head()
[52]:
                    location
                               size
                                     total_sqft bath price
                                                              bhk price_per_sqft
      0 1st Block Jayanagar 4 BHK
                                         2850.0
                                                  4.0 428.0
                                                                     15017.543860
      1 1st Block Jayanagar
                             3 BHK
                                         1630.0
                                                  3.0 194.0
                                                                3
                                                                     11901.840491
      2 1st Block Jayanagar 3 BHK
                                         1875.0
                                                  2.0 235.0
                                                                     12533.333333
      3 1st Block Jayanagar 3 BHK
                                         1200.0
                                                  2.0 130.0
                                                                3
                                                                     10833.333333
      4 1st Block Jayanagar 2 BHK
                                         1235.0
                                                  2.0 148.0
                                                                     11983.805668
[53]: data.drop(columns = ['size', 'price_per_sqft'], inplace=True)
      # price_per_sqft only used to extract the outliers from the data
[54]: data.head()
[54]:
                    location total_sqft bath price bhk
      0 1st Block Jayanagar
                                           4.0 428.0
                                  2850.0
      1 1st Block Jayanagar
                                  1630.0
                                           3.0 194.0
                                                         3
      2 1st Block Jayanagar
                                  1875.0
                                           2.0 235.0
      3 1st Block Jayanagar
                                  1200.0
                                           2.0 130.0
                                                         3
      4 1st Block Jayanagar
                                  1235.0
                                           2.0 148.0
                                                         2
[55]: data.to_csv("Cleaned_data.csv")
[56]: X = data.drop(columns = ['price'])
                                           # X without price
      y = data['price']
                          # y is only price
[57]: X.head()
```

```
[57]:
                   location total_sqft bath bhk
     0 1st Block Jayanagar
                                 2850.0
                                          4.0
                                                 4
     1 1st Block Jayanagar
                                          3.0
                                 1630.0
                                                 3
     2 1st Block Jayanagar
                                 1875.0
                                          2.0
                                                 3
     3 1st Block Jayanagar
                                 1200.0
                                          2.0
                                                 3
     4 1st Block Jayanagar
                                 1235.0
                                          2.0
                                                 2
[58]: from sklearn.model_selection import train_test_split
     from sklearn.linear_model import LinearRegression, Lasso, Ridge
     from sklearn.preprocessing import OneHotEncoder, StandardScaler
     from sklearn.compose import make_column_transformer
     from sklearn.pipeline import make_pipeline
     from sklearn.metrics import r2_score
[59]: X.shape
[59]: (7361, 4)
[60]: X_train, X_test, y_train, y_test = train_test_split(X,y, test_size = 0.2,__
       →random state=0)
[61]: print(X_train.shape)
     print(X_test.shape)
      # 5888+1473 = 7361
      # also 20% of 7361 is 1473
     (5888, 4)
     (1473, 4)
[62]: column_trans = make_column_transformer((OneHotEncoder(sparse_output=False),_
       [63]: scaler = StandardScaler()
        Linear Regression
[64]: pipe = make_pipeline(column_trans, scaler, LinearRegression())
[65]: pipe.fit(X_train, y_train)
[65]: Pipeline(steps=[('columntransformer',
                      ColumnTransformer(remainder='passthrough',
                                        transformers=[('onehotencoder',
     OneHotEncoder(sparse_output=False),
                                                       ['location'])])),
                     ('standardscaler', StandardScaler()),
                     ('linearregression', LinearRegression())])
```

```
[66]: y_pred_lr = pipe.predict(X_test)
[67]: r2_score(y_test, y_pred_lr)
[67]: 0.825224489680086
        Lasso
[68]: lasso = Lasso()
[69]: pipe = make_pipeline(column_trans, scaler, lasso)
[70]: pipe.fit(X_train, y_train)
[70]: Pipeline(steps=[('columntransformer',
                       ColumnTransformer(remainder='passthrough',
                                         transformers=[('onehotencoder',
      OneHotEncoder(sparse_output=False),
                                                         ['location'])])),
                      ('standardscaler', StandardScaler()), ('lasso', Lasso())])
[71]: | y_pred_lasso = pipe.predict(X_test)
[72]: r2_score(y_test, y_pred_lasso)
[72]: 0.814689475169039
         Ridge
[73]: ridge = Ridge()
[74]: pipe = make_pipeline(column_trans, scaler, ridge)
[75]: pipe.fit(X_train, y_train)
[75]: Pipeline(steps=[('columntransformer',
                       ColumnTransformer(remainder='passthrough',
                                         transformers=[('onehotencoder',
      OneHotEncoder(sparse_output=False),
                                                         ['location'])])),
                      ('standardscaler', StandardScaler()), ('ridge', Ridge())])
[76]: y_pred_ridge = pipe.predict(X_test)
[77]: r2_score(y_test, y_pred_ridge)
[77]: 0.8252348502290107
```

```
[78]: print('No Regularization:', r2_score(y_test, y_pred_lr))
    print('Lasso:', r2_score(y_test, y_pred_lasso))
    print('Ridge:', r2_score(y_test, y_pred_ridge))

No Regularization: 0.825224489680086
    Lasso: 0.814689475169039
    Ridge: 0.8252348502290107

[79]: import pickle

[80]: pickle.dump(pipe, open('RidgeModel.pkl','wb'))
```

7 Thank You by Navjoth Singh