Data Dictionary

Variable Name	Description
ID	Identifier ID
title	Title from listing
subtitle	Neighborhood and city
sq_mt_built	Square meter built
sq_mt_useful	Square meter useful
n_rooms	Number of rooms
n_bathrooms	Number of bathrooms
sq_mt_allotment	Square meter allotment
latitude, longitude	Latitude, Longitude (Since portals hide the location mainly NAs)
raw_address	Address
is_exact_address_hidden	Boolean values
buy_price	Target Value

```
import pandas as pd
import numpy as np
import plotly.express as px
from matplotlib import pyplot as plt
import seaborn as sns

pd.options.display.float_format = '{:,.2f}'.format
pd.options.display.max_rows = None
pd.options.display.max_columns = None
```

```
sns.set()
In [2]: data = pd.read_csv('houses_Madrid.csv')
In [3]: data.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 21742 entries, 0 to 21741
Data columns (total 58 columns):

Data	columns (total 58 columns):		
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	21742 non-null	int64
1	id	21742 non-null	int64
2	title	21742 non-null	object
3	subtitle	21742 non-null	object
4	sq_mt_built	21616 non-null	float64
	· - -	8228 non-null	float64
5	sq_mt_useful		
6	n_rooms	21742 non-null	int64
7	n_bathrooms	21726 non-null	float64
8	n_floors	1437 non-null	float64
9	sq_mt_allotment	1432 non-null	float64
10	latitude	0 non-null	float64
11	longitude	0 non-null	float64
12	raw_address	16277 non-null	object
13	is_exact_address_hidden	21742 non-null	bool
14	street_name	15837 non-null	object
15	street_number	6300 non-null	object
16	portal	0 non-null	float64
17	floor	19135 non-null	object
18	is_floor_under	20572 non-null	object
19	door	0 non-null	float64
20	neighborhood_id	21742 non-null	object
21	operation	21742 non-null	object
22	rent_price	21742 non-null	int64
23	rent_price_by_area	0 non-null	float64
24	is_rent_price_known	21742 non-null	bool
25	buy_price	21742 non-null	int64
26	buy_price_by_area	21742 non-null	int64
27	is_buy_price_known	21742 non-null	bool
28	house_type_id	21351 non-null	object
29	is_renewal_needed	21742 non-null	bool
30	is new development	20750 non-null	object
31	built_year	10000 non-null	•
32	has_central_heating	13608 non-null	
33	has_individual_heating	13608 non-null	object
34	are_pets_allowed	0 non-null	float64
35	has_ac	11211 non-null	object
36	has_fitted_wardrobes	13399 non-null	object
37		19356 non-null	•
	has_lift	18699 non-null	object
38	is_exterior		object
39	has_garden	1556 non-null	object
40	has_pool	5171 non-null	object
41	has_terrace	9548 non-null	object
42	has_balcony	3321 non-null	object
43	has_storage_room	7698 non-null	object
44	is_furnished	0 non-null	float64
45	is_kitchen_equipped	0 non-null	float64
46	is_accessible	4074 non-null	object
47	has_green_zones	4057 non-null	object
48	energy_certificate	21742 non-null	object
49	has_parking	21742 non-null	bool
50	has_private_parking	0 non-null	float64

```
51 has_public_parking
                               0 non-null
                                             float64
52 is_parking_included_in_price 7719 non-null object
53 parking_price
                               7719 non-null float64
54 is_orientation_north
                             11358 non-null object
55 is_orientation_west
                               11358 non-null object
56 is_orientation_south
                               11358 non-null object
                           11358 non-null object
57 is_orientation_east
dtypes: bool(5), float64(17), int64(6), object(30)
memory usage: 8.9+ MB
```

```
In [4]: data = data.dropna(axis=1, how='all') #remove column(axis=1), where all values are
data.info()
```

	columns (total 48 columns):		
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	21742 non-null	int64
1	id	21742 non-null	int64
2	title	21742 non-null	object
3	subtitle	21742 non-null	object
4	sq_mt_built	21616 non-null	float64
5	sq_mt_useful	8228 non-null	float64
6	n_rooms	21742 non-null	int64
7	n_bathrooms	21726 non-null	float64
8	n_floors	1437 non-null	float64
9	sq_mt_allotment	1432 non-null	float64
10	raw_address	16277 non-null	•
11	is_exact_address_hidden	21742 non-null	bool
12	street_name	15837 non-null	object
13	street_number	6300 non-null	object
14	floor	19135 non-null	object
15	is_floor_under	20572 non-null	object
16	neighborhood_id	21742 non-null	object
17	operation	21742 non-null	object
18	rent_price	21742 non-null	int64
19	is_rent_price_known	21742 non-null	bool
20	buy_price	21742 non-null	int64
21	buy_price_by_area	21742 non-null	int64
22	is_buy_price_known	21742 non-null	bool
23	house_type_id	21351 non-null	object
24	is_renewal_needed	21742 non-null	bool
25	is_new_development	20750 non-null	object
26	built_year	10000 non-null	float64
27	has_central_heating	13608 non-null	object
28	has_individual_heating	13608 non-null	object
29	has_ac	11211 non-null	object
30	has_fitted_wardrobes	13399 non-null	object
31	has_lift	19356 non-null	object
32	is_exterior	18699 non-null	object
33	has_garden	1556 non-null	object
34	has_pool	5171 non-null	object
35	has_terrace	9548 non-null	object
36	has_balcony	3321 non-null	object
37	has_storage_room	7698 non-null	object
38	is_accessible	4074 non-null	object
39	has_green_zones	4057 non-null	object
40	energy_certificate	21742 non-null	object
41	has_parking	21742 non-null	bool
42	<pre>is_parking_included_in_price</pre>	7719 non-null	object
43	parking_price	7719 non-null	float64
44	is_orientation_north	11358 non-null	object
45	is_orientation_west	11358 non-null	object
46	is_orientation_south	11358 non-null	object
47	is_orientation_east	11358 non-null	object
	 es: bool(5), float64(7), int64		•
	ον μεραρ: 7 2± MR		

memory usage: 7.2+ MB

```
In [5]: data.duplicated().sum()
Out[5]: np.int64(0)
In [6]: df = data[['id', 'subtitle', 'sq_mt_built', 'sq_mt_useful', 'n_rooms', 'n_bathrooms
        df.info()
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 21742 entries, 0 to 21741
      Data columns (total 13 columns):
           Column
                             Non-Null Count Dtype
           _____
                             -----
       0
           id
                             21742 non-null int64
                             21742 non-null object
       1
           subtitle
       2
           sq_mt_built
                             21616 non-null float64
       3
                             8228 non-null float64
           sq_mt_useful
                             21742 non-null int64
       4
           n rooms
                             21726 non-null float64
       5
           n_bathrooms
       6
           n_floors
                             1437 non-null float64
       7
           sq_mt_allotment 1432 non-null float64
       8
           rent_price
                            21742 non-null int64
       9
           buy_price
                             21742 non-null int64
       10 buy_price_by_area 21742 non-null int64
       11 built_year
                             10000 non-null float64
       12 parking_price
                             7719 non-null float64
      dtypes: float64(7), int64(5), object(1)
      memory usage: 2.2+ MB
        df=df.rename(columns={'subtitle' : 'city'})
In [8]: | df = df[['id', 'city', 'sq_mt_built', 'n_rooms', 'n_bathrooms', 'rent_price', 'buy_
        df.info()
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 21742 entries, 0 to 21741
      Data columns (total 8 columns):
       # Column
                             Non-Null Count Dtype
       --- -----
                             -----
       0
           id
                             21742 non-null int64
                             21742 non-null object
       1
           city
                             21616 non-null float64
       2
           sq_mt_built
       3
                             21742 non-null int64
           n_rooms
       4
           n_bathrooms
                             21726 non-null float64
       5
                             21742 non-null int64
           rent_price
           buy_price
                             21742 non-null int64
       7
           buy_price_by_area 21742 non-null int64
      dtypes: float64(2), int64(5), object(1)
      memory usage: 1.3+ MB
In [9]: df.n_rooms.describe()
```

```
Out[9]: count
                 21,742.00
         mean
                      3.01
                      1.51
         std
         min
                      0.00
         25%
                      2.00
         50%
                      3.00
         75%
                      4.00
                     24.00
         max
         Name: n_rooms, dtype: float64
In [10]: df.drop(df.index[df.n rooms == 0], axis=0, inplace=True)
In [11]: df.n_bathrooms.mode()
Out[11]: 0 1.00
         Name: n_bathrooms, dtype: float64
In [12]: df.n bathrooms.fillna(1, inplace=True)
         df.n_bathrooms = df.n_bathrooms.replace(0,1)
        C:\Users\alhef\AppData\Local\Temp\ipykernel_7008\3396593470.py:1: FutureWarning: A v
        alue is trying to be set on a copy of a DataFrame or Series through chained assignme
        nt using an inplace method.
        The behavior will change in pandas 3.0. This inplace method will never work because
        the intermediate object on which we are setting values always behaves as a copy.
        For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method
        ({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform
        the operation inplace on the original object.
          df.n_bathrooms.fillna(1, inplace=True)
In [13]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        Index: 21303 entries, 0 to 21741
        Data columns (total 8 columns):
                               Non-Null Count Dtype
            Column
        ---
            -----
                                21303 non-null int64
         0
             id
         1
            city
                               21303 non-null object
         2
             sq_mt_built
                               21177 non-null float64
         3
            n_rooms
                               21303 non-null int64
        4
             n_bathrooms
                               21303 non-null float64
                               21303 non-null int64
         5
             rent price
                               21303 non-null int64
             buy_price
             buy_price_by_area 21303 non-null int64
         7
        dtypes: float64(2), int64(5), object(1)
        memory usage: 1.5+ MB
In [14]: df.sample(10)
```

```
Out[14]:
                     id
                                   city sq_mt_built n_rooms n_bathrooms rent_price buy_price
                              Malasaña-
                                                            1
          18259
                   3483
                            Universidad,
                                              43.00
                                                                       1.00
                                                                                  936
                                                                                          206000
                                Madrid
                         Palomeras Bajas,
                16968
                                               53.00
                                                            1
                                                                       1.00
                                                                                  621
                                                                                          119000
           4774
                                Madrid
                            Almendrales,
           2254 19488
                                              102.00
                                                            2
                                                                       1.00
                                                                                   508
                                                                                           93000
                                Madrid
                         Chueca-Justicia,
          21630
                   112
                                              45.00
                                                            1
                                                                       1.00
                                                                                 1275
                                                                                          330000
                                Madrid
           6328
                 15414
                         Águilas, Madrid
                                              80.00
                                                            3
                                                                       2.00
                                                                                  964
                                                                                          215000
          12923
                  8819
                                                                       3.00
                                                                                 2378
                         Atalaya, Madrid
                                              169.00
                                                                                          825000
           9143 12599
                            Pilar, Madrid
                                              54.00
                                                            1
                                                                       1.00
                                                                                  851
                                                                                          180000
                             Fuencarral,
           9292
                12450
                                              74.00
                                                            3
                                                                       1.00
                                                                                  837
                                                                                          176000
                                Madrid
           6650 15092
                          Lucero, Madrid
                                                                       1.00
                                              48.00
                                                            1
                                                                                  625
                                                                                          119900
                              Bernabéu-
          14058
                  7684
                        Hispanoamérica,
                                              330.00
                                                            5
                                                                       4.00
                                                                                 -2214
                                                                                         1495000
                                Madrid
         df.n_bathrooms = df.n_bathrooms.astype(int)
In [15]:
In [16]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        Index: 21303 entries, 0 to 21741
        Data columns (total 8 columns):
             Column
                                  Non-Null Count Dtype
             ____
                                  _____
         0
              id
                                  21303 non-null int64
         1
                                  21303 non-null object
              city
                                  21177 non-null float64
         2
              sq_mt_built
         3
             n_rooms
                                  21303 non-null int64
         4
              n_bathrooms
                                  21303 non-null int64
         5
              rent_price
                                  21303 non-null int64
              buy_price
         6
                                  21303 non-null int64
              buy_price_by_area 21303 non-null int64
        dtypes: float64(1), int64(6), object(1)
        memory usage: 1.5+ MB
In [17]: #drop sq_mt_built NaN rows
          df.dropna(subset='sq_mt_built', inplace=True)
```

In [18]: df.info()

<class 'pandas.core.frame.DataFrame'> Index: 21177 entries, 0 to 21741 Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	id	21177 non-null	int64
1	city	21177 non-null	object
2	sq_mt_built	21177 non-null	float64
3	n_rooms	21177 non-null	int64
4	n_bathrooms	21177 non-null	int64
5	rent_price	21177 non-null	int64
6	buy_price	21177 non-null	int64
7	<pre>buy_price_by_area</pre>	21177 non-null	int64

dtypes: float64(1), int64(6), object(1)

memory usage: 1.5+ MB

In [19]: df.sample(10)

Out[19]:

	id	city	sq_mt_built	n_rooms	n_bathrooms	rent_price	buy_price	bu
1234	20508	Vicálvaro, Madrid	88.00	3	2	1174	290000	
13403	8339	Ventas, Madrid	62.00	2	1	666	130000	
8915	12827	Las Tablas, Madrid	93.00	3	2	1644	490000	
570	21172	Los Ángeles, Madrid	70.00	3	1	621	119000	
17298	4444	Lavapiés- Embajadores, Madrid	57.00	1	1	1035	239000	
5572	16170	Numancia, Madrid	65.00	2	1	662	129000	
6031	15711	Los Cármenes, Madrid	80.00	3	1	686	135000	
76	21666	Butarque, Madrid	117.00	3	2	1052	244798	
534	21208	San Cristóbal, Madrid	62.00	3	1	574	108000	
20869	873	Legazpi, Madrid	140.00	3	2	1428	395000	
4								•

In [20]: #change negative values in Rent price column to Zero for value in df.rent_price.values:

```
if value < 0:</pre>
                 df.rent_price = 0
In [21]: df[df.rent_price < 0]</pre>
Out[21]:
           id city sq_mt_built n_rooms n_bathrooms rent_price buy_price_by_area
In [22]: #split City Column to (district, city)
         df['district'] = df.city.str.split(',', expand=True)[0]
In [23]: df.head()
Out[23]:
               id
                       city sq_mt_built n_rooms n_bathrooms rent_price buy_price_b
                        San
         0 21742 Cristóbal,
                                  64.00
                                              2
                                                           1
                                                                      0
                                                                            85000
                     Madrid
                        Los
                                              3
                                                                      0
          1 21741
                                  70.00
                                                            1
                    Ángeles,
                                                                           129900
                     Madrid
                       San
                                              2
                                                            2
                                                                           144247
         2 21740
                     Andrés,
                                  94.00
                                                                      0
                     Madrid
                        San
         3 21739
                     Andrés,
                                  64.00
                                              2
                                                            1
                                                                      0
                                                                           109900
                     Madrid
                        Los
         4 21738
                                 108.00
                                              2
                                                            2
                                                                      0
                                                                           260000
                    Rosales,
                     Madrid
In [24]: df = df[['id', 'district', 'sq_mt_built', 'n_rooms', 'n_bathrooms', 'rent_price',
                 'buy_price', 'buy_price_by_area']]
         df.head()
```

Out[24]:		id	district	sq_mt_built	n_rooms	n_bathrooms	rent_price	buy_price	buy_price_b
	0	21742	San Cristóbal	64.00	2	1	0	85000	
	1	21741	Los Ángeles	70.00	3	1	0	129900	
	2	21740	San Andrés	94.00	2	2	0	144247	
	3	21739	San Andrés	64.00	2	1	0	109900	
	4	21738	Los Rosales	108.00	2	2	0	260000	
	4								—
In [25]:	da	ta.head	()						
Out[25]:		Unnam	ned: 0 i	d title	subtitle	sq_mt_built	sq_mt_useful	n_rooms	n_bathroom
	0		0 2174	Piso en venta en 2 calle de Godella, 64	San Cristóbal, Madrid	64.00	60.00	2	1.0
	1		1 2174	Piso en venta en calle de la del Manojo de Rosas	Los Ángeles, Madrid	70.00	NaN	3	1.0
	2		2 2174	Piso en venta en calle del Talco, 68	San Andrés, Madrid	94.00	54.00	2	2.0
	3		3 2173	Piso en venta en 9 calle Pedro Jiménez	San Andrés, Madrid	64.00	NaN	2	1.0

Los

108.00

Rosales,

Madrid

2

90.00

2.0

Piso en venta en

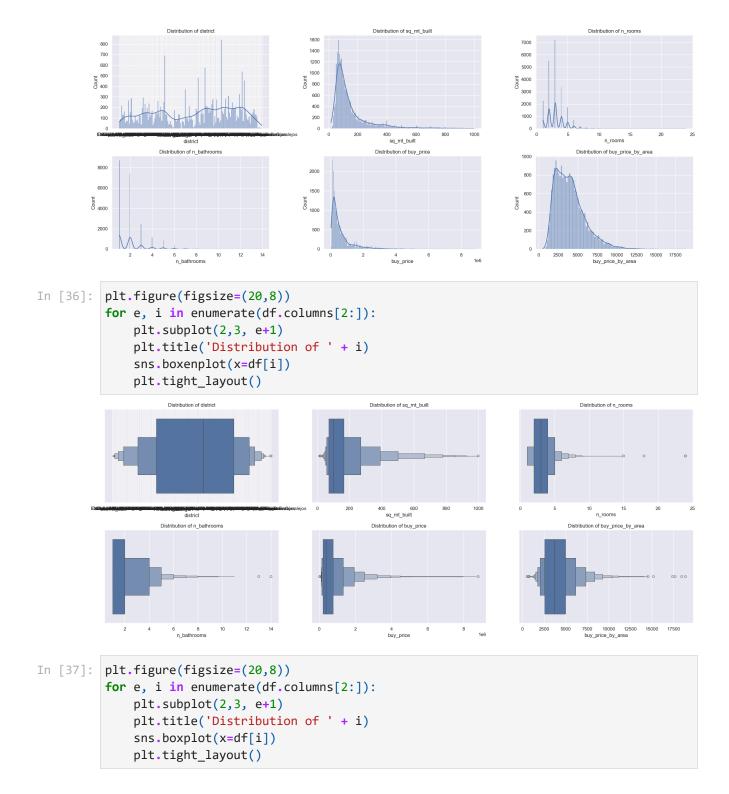
carretera

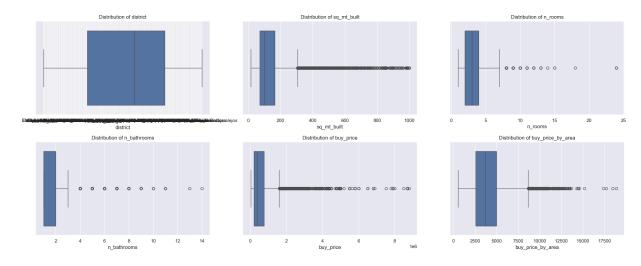
Villaverde a Val...

de

4 21738

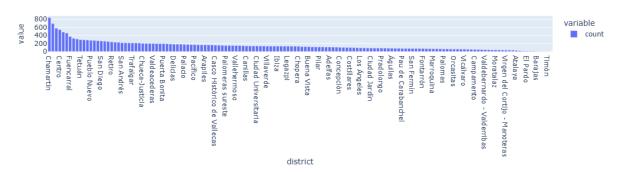
```
<class 'pandas.core.frame.DataFrame'>
       RangeIndex: 21742 entries, 0 to 21741
       Data columns (total 3 columns):
        # Column
                                    Non-Null Count Dtype
       --- -----
                                    -----
        0 raw_address
                                   16277 non-null object
            is_exact_address_hidden 21742 non-null bool
                                   15837 non-null object
            street_name
       dtypes: bool(1), object(2)
       memory usage: 361.1+ KB
In [29]: data['operation'].unique()
Out[29]: array(['sale'], dtype=object)
In [30]: # Remove rent_price
         df.drop(columns=['rent_price'], inplace=True)
In [31]: df.to_csv('Madrid_cleaned_data.csv')
In [32]: df = pd.read_csv('Madrid_cleaned_data.csv')
In [33]: df.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 21177 entries, 0 to 21176
       Data columns (total 8 columns):
        # Column
                              Non-Null Count Dtype
       ---
                             -----
                            21177 non-null int64
        0
           Unnamed: 0
        1
                             21177 non-null int64
           id
                           21177 non-null object
21177 non-null float64
           district
        2
        3
           sq_mt_built
        4 n_rooms
                            21177 non-null int64
                            21177 non-null int64
        5
           n_bathrooms
                             21177 non-null int64
            buy_price
            buy_price_by_area 21177 non-null int64
        7
       dtypes: float64(1), int64(6), object(1)
       memory usage: 1.3+ MB
In [34]: plt.figure(figsize=(20,8))
         for e, i in enumerate(df.columns[2:]):
            plt.subplot(2,3, e+1)
            plt.title('Distribution of ' + i)
            sns.histplot(df[i],kde=True)
            plt.tight_layout()
```





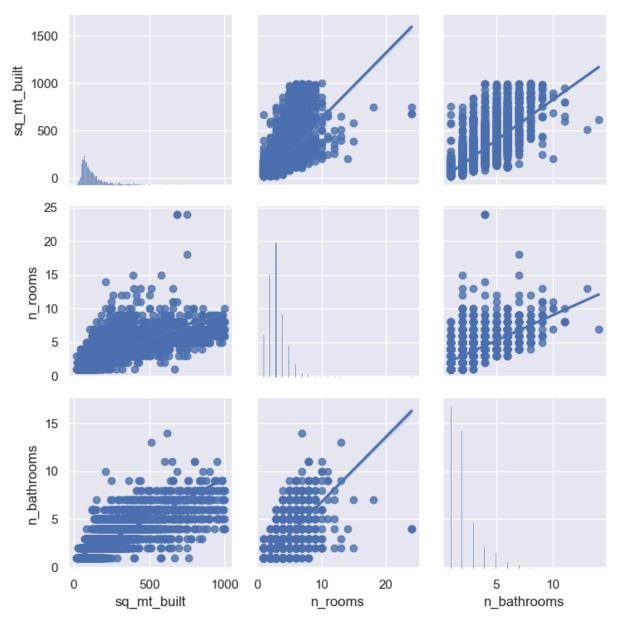
In [39]: px.bar(df.district.value_counts(), title= 'Distribution Of Districts in Madrid')

Distribution Of Districts in Madrid



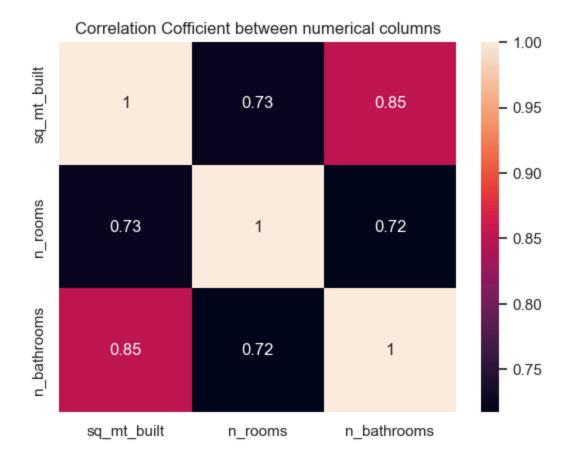
Correllation Coeffcient

```
In [40]: corr = df[['sq_mt_built', 'n_rooms', 'n_bathrooms']]
    sns.pairplot(corr,kind='reg')
    plt.tight_layout()
```



In [41]: sns.heatmap(corr.corr(), annot=True)
 plt.title('Correlation Cofficient between numerical columns')

Out[41]: Text(0.5, 1.0, 'Correlation Cofficient between numerical columns')



```
In [42]: sns.regplot(data=df, x='sq_mt_built', y='buy_price')
  plt.title('Correlation between Area and Buy Price')
  plt.xlabel('Area')
  plt.ylabel('Buy Price')
```

Out[42]: Text(0, 0.5, 'Buy Price')



In [43]: df.head()

ut[43]:		Unnamed:	id	district	sq_mt_built	n_rooms	n_bathrooms	buy_price	buy_price_b
	0	0	21742	San Cristóbal	64.00	2	1	85000	
	1	1	21741	Los Ángeles	70.00	3	1	129900	
	2	2	21740	San Andrés	94.00	2	2	144247	
	3	3	21739	San Andrés	64.00	2	1	109900	
	4	4	21738	Los Rosales	108.00	2	2	260000	
	4								

In [46]: df.drop('Unnamed: 0', axis=1, inplace=True)

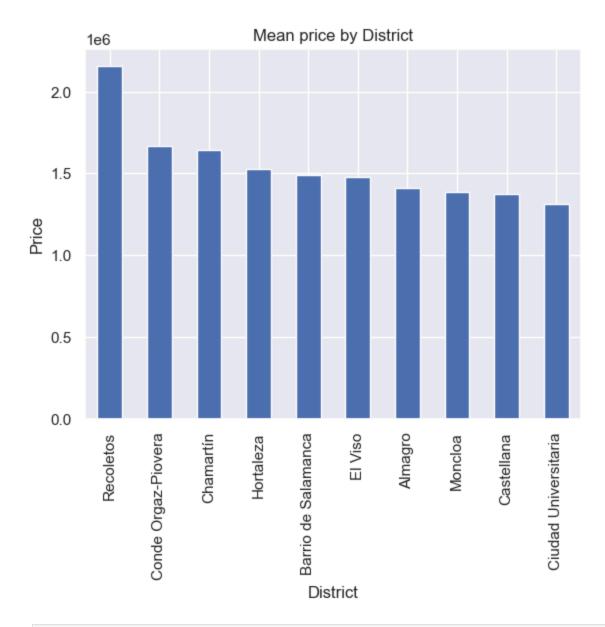
In [47]: df.head()

Out[47]:	ut[47]: id di		district	sq_mt_built	n_rooms	n_bathrooms	buy_price	buy_price_by_area
	0	21742	San Cristóbal	64.00	2	1	85000	1328
	1	21741	Los Ángeles	70.00	3	1	129900	1856
	2	21740	San Andrés	94.00	2	2	144247	1535
	3	21739	San Andrés	64.00	2	1	109900	1717
	4	21738	Los Rosales	108.00	2	2	260000	2407

Location or Size : What influence House prices in Madrid?

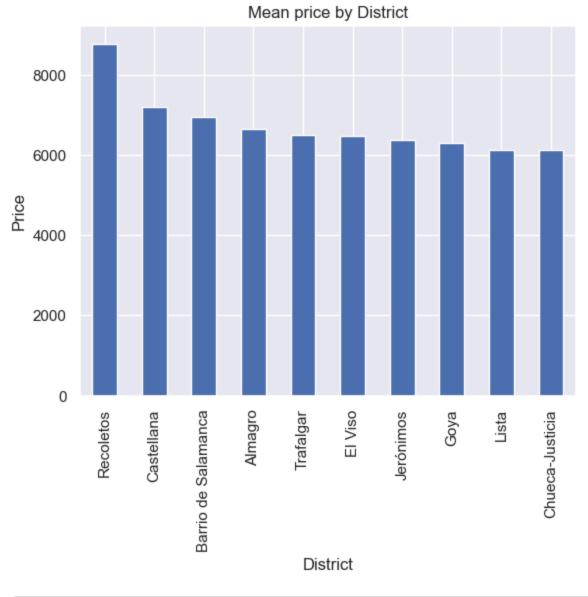
```
In [50]: mean_price = df.groupby('district')['buy_price'].mean().sort_values(ascending=False
    mean_price.head(10).plot(kind='bar', xlabel='District', ylabel = 'Price', title='Me

Out[50]: <Axes: title={'center': 'Mean price by District'}, xlabel='District', ylabel='Price', vlabel='Price')</pre>
```



```
In [51]: mean_price = df.groupby('district')['buy_price_by_area'].mean().sort_values(ascendi
mean_price.head(10).plot(kind='bar', xlabel='District', ylabel = 'Price', title='Me
```

Out[51]: <Axes: title={'center': 'Mean price by District'}, xlabel='District', ylabel='Pric
 e'>



```
In [53]: reco = df.query('district == "Recoletos"')
In [54]: reco.sample(10)
```

Out[54]:		id	district	sq_mt_built	n_rooms	n_bathrooms	buy_price	buy_price_by_area
	18917	2297	Recoletos	404.00	5	6	4350000	10767
	18596	2620	Recoletos	92.00	2	1	378000	4109
	18962	2250	Recoletos	210.00	6	3	1800000	8571
	19168	2043	Recoletos	410.00	5	3	3300000	8049
	19412	1795	Recoletos	404.00	5	6	4350000	10767
	19160	2051	Recoletos	247.00	2	3	2200000	8907
	19171	2040	Recoletos	706.00	6	7	6500000	9207
	18809	2406	Recoletos	471.00	6	6	4155000	8822
	18766	2449	Recoletos	365.00	4	5	3285000	9000
	19137	2074	Recoletos	70.00	1	1	568000	8114

In []:

In [56]: pip install statsmodels

```
Collecting statsmodels
 Downloading statsmodels-0.14.5-cp313-cp313-win_amd64.whl.metadata (9.8 kB)
Requirement already satisfied: numpy<3,>=1.22.3 in c:\users\alhef\appdata\local\prog
rams\python\python313\lib\site-packages (from statsmodels) (2.3.2)
Requirement already satisfied: scipy!=1.9.2,>=1.8 in c:\users\alhef\appdata\local\pr
ograms\python\python313\lib\site-packages (from statsmodels) (1.16.1)
Requirement already satisfied: pandas!=2.1.0,>=1.4 in c:\users\alhef\appdata\local\p
rograms\python\python313\lib\site-packages (from statsmodels) (2.3.1)
Collecting patsy>=0.5.6 (from statsmodels)
 Downloading patsy-1.0.1-py2.py3-none-any.whl.metadata (3.3 kB)
Requirement already satisfied: packaging>=21.3 in c:\users\alhef\appdata\local\progr
ams\python\python313\lib\site-packages (from statsmodels) (25.0)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\alhef\appdata\loca
l\programs\python\python313\lib\site-packages (from pandas!=2.1.0,>=1.4->statsmodel
s) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\alhef\appdata\local\programs
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Requirement already satisfied: tzdata>=2022.7 in c:\users\alhef\appdata\local\progra
ms\python\python313\lib\site-packages (from pandas!=2.1.0,>=1.4->statsmodels) (2025.
Requirement already satisfied: six>=1.5 in c:\users\alhef\appdata\local\programs\pyt
hon\python313\lib\site-packages (from python-dateutil>=2.8.2->pandas!=2.1.0,>=1.4->s
tatsmodels) (1.17.0)
Downloading statsmodels-0.14.5-cp313-cp313-win_amd64.whl (9.6 MB)
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  ----- 7.6/9.6 MB 3.5 MB/s eta 0:00:01
  ----- 8.4/9.6 MB 3.5 MB/s eta 0:00:01
  ----- 9.2/9.6 MB 3.5 MB/s eta 0:00:01
  ----- 9.6/9.6 MB 3.4 MB/s 0:00:02
Downloading patsy-1.0.1-py2.py3-none-any.whl (232 kB)
Installing collected packages: patsy, statsmodels
  ----- 0/2 [patsy]
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```
----- 1/2 [statsmodels]
----- 2/2 [statsmodels]
```

Successfully installed patsy-1.0.1 statsmodels-0.14.5 Note: you may need to restart the kernel to use updated packages.

```
In [58]: import statsmodels.api as sm
  reco['intercept'] = 1
  lm = sm.OLS(reco['buy_price'], reco[['intercept', 'sq_mt_built']])
  result = lm.fit()
  result.summary()
```

C:\Users\alhef\AppData\Local\Temp\ipykernel_7008\2760881175.py:2: SettingWithCopyWar
ning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

OLS Regression Results

Dep. Variable:	buy_price	R-squared:	0.833
Model:	OLS	Adj. R-squared:	0.832
Method:	Least Squares	F-statistic:	555.1
Date:	Sat, 16 Aug 2025	Prob (F-statistic):	5.30e-45
Time:	14:14:00	Log-Likelihood:	-1654.1
No. Observations:	113	AIC:	3312.
Df Residuals:	111	BIC:	3318.
Df Model:	1		

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
intercept	-4.342e+04	1.07e+05	-0.406	0.686	-2.55e+05	1.69e+05
sq_mt_built	8975.6204	380.945	23.561	0.000	8220.753	9730.488

Omnibus:	11.498	Durbin-Watson:	2.039
Prob(Omnibus):	0.003	Jarque-Bera (JB):	16.048
Skew:	0.512	Prob(JB):	0.000328
Kurtosis:	4.536	Cond. No.	574.

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [59]: x = 300
#YHat = b0 + b1 *x
price = 152500 + 6843*x
print("Predicted price of flat in Recoletos with area of {} m2 = {} Euro".format(x,)
Predicted price of flat in Recoletos with area of 300 m2 = 2205400 Euro
```