Panda's assignment

December 14, 2017

In [1]:
```
### Name:Thinley Jurmi ID:u6018302
```

In [ ]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt #importing pandas
df=pd.read_csv("/Users/thinleyjurmi/Downloads/untitled folder/CArealestate.csv")
#csv used
```

In [2]:
```
df.head(20)#first 20 data of the csv file
```

Out[2]:

	street	city	zip	state	beds	baths
0	3526 HIGH ST	SACRAMENTO	95838	CA	2	1
1	51 OMAHA CT	SACRAMENTO	95823	CA	3	1
2	2796 BRANCH ST	SACRAMENTO	95815	CA	2	1
3	2805 JANETTE WAY	SACRAMENTO	95815	CA	2	1
4	6001 MCMAHON DR	SACRAMENTO	95824	CA	2	1
5	5828 PEPPERMILL CT	SACRAMENTO	95841	CA	3	1
6	6048 OGDEN NASH WAY	SACRAMENTO	95842	CA	3	2
7	2561 19TH AVE	SACRAMENTO	95820	CA	3	1
8	11150 TRINITY RIVER DR Unit 114	RANCHO CORDOVA	95670	CA	2	2
9	7325 10TH ST	RIO LINDA	95673	CA	3	2
10	645 MORRISON AVE	SACRAMENTO	95838	CA	3	2
11	4085 FAWN CIR	SACRAMENTO	95823	CA	3	2
12	2930 LA ROSA RD	SACRAMENTO	95815	CA	1	1
13	2113 KIRK WAY	SACRAMENTO	95822	CA	3	1
14	4533 LOCH HAVEN WAY	SACRAMENTO	95842	CA	2	2
15	7340 HAMDEN PL	SACRAMENTO	95842	CA	2	2
16	6715 6TH ST	RIO LINDA	95673	CA	2	1
17	6236 LONGFORD DR Unit 1	CITRUS HEIGHTS	95621	CA	2	1
18	250 PERALTA AVE	SACRAMENTO	95833	CA	2	1
19	113 LEEWILL AVE	RIO LINDA	95673	CA	3	2

	sq__ft	type	sale_date	price	latitude
0	836	Residential	Wed May 21 00:00:00 EDT 2008	59222	38.631913
1	1167	Residential	Wed May 21 00:00:00 EDT 2008	68212	38.478902
2	796	Residential	Wed May 21 00:00:00 EDT 2008	68880	38.618305
3	852	Residential	Wed May 21 00:00:00 EDT 2008	69307	38.616835
4	797	Residential	Wed May 21 00:00:00 EDT 2008	81900	38.519470

```
5     1122        Condo  Wed May 21 00:00:00 EDT 2008   89921  38.662595
6     1104  Residential  Wed May 21 00:00:00 EDT 2008   90895  38.681659
7     1177  Residential  Wed May 21 00:00:00 EDT 2008   91002  38.535092
8      941        Condo  Wed May 21 00:00:00 EDT 2008   94905  38.621188
9     1146  Residential  Wed May 21 00:00:00 EDT 2008   98937  38.700909
10     909  Residential  Wed May 21 00:00:00 EDT 2008  100309  38.637663
11    1289  Residential  Wed May 21 00:00:00 EDT 2008  106250  38.470746
12     871  Residential  Wed May 21 00:00:00 EDT 2008  106852  38.618698
13    1020  Residential  Wed May 21 00:00:00 EDT 2008  107502  38.482215
14    1022  Residential  Wed May 21 00:00:00 EDT 2008  108750  38.672914
15    1134        Condo  Wed May 21 00:00:00 EDT 2008  110700  38.700051
16     844  Residential  Wed May 21 00:00:00 EDT 2008  113263  38.689591
17     795        Condo  Wed May 21 00:00:00 EDT 2008  116250  38.679776
18     588  Residential  Wed May 21 00:00:00 EDT 2008  120000  38.612099
19    1356  Residential  Wed May 21 00:00:00 EDT 2008  121630  38.689999

     longitude
0  -121.434879
1  -121.431028
2  -121.443839
3  -121.439146
4  -121.435768
5  -121.327813
6  -121.351705
7  -121.481367
8  -121.270555
9  -121.442979
10 -121.451520
11 -121.458918
12 -121.435833
13 -121.492603
14 -121.359340
15 -121.351278
16 -121.452239
17 -121.314089
18 -121.469095
19 -121.463220
```

In [3]:
```
df.describe()
```

Out[3]:
```
                zip        beds       baths       sq__ft          price  \
count    499.000000  499.000000  499.000000   499.000000     499.000000
mean   95758.643287    3.062124    1.871743  1410.086172  232728.452906
std       84.660893    1.145161    0.838089   775.033290  121169.372390
min    95603.000000    0.000000    0.000000     0.000000   40000.000000
25%    95670.000000    3.000000    1.000000  1019.000000  145500.000000
50%    95815.000000    3.000000    2.000000  1326.000000  215100.000000
75%    95828.000000    4.000000    2.000000  1776.000000  285000.000000
```

```
max    95864.000000    8.000000    5.000000  3992.000000  839000.000000
```

```
          latitude   longitude
count  499.000000  499.000000
mean    38.580031 -121.369985
std      0.129570    0.132910
min     38.241514 -121.551704
25%     38.475161 -121.450145
50%     38.602416 -121.401482
75%     38.676239 -121.315538
max     39.008159 -120.603872
```

In [6]:
```
df.groupby(["city", "street"]).size()#street grouped to a city accordingly
```

Out[6]:
```
city          street
ANTELOPE      3228 BAGGAN CT            1
              3305 RIO ROCA CT          1
              3318 DAVIDSON DR          1
              3828 BLACKFOOT WAY        1
              4437 MITCHUM CT           1
              4508 OLD DAIRY DR         1
              4844 CLYDEBANK WAY        1
              5308 MARBURY WAY          1
              5312 MARBURY WAY          1
              5708 RIDGEPOINT DR        1
              5712 MELBURY CIR          1
              7837 ABBINGTON WAY        1
              7863 CRESTLEIGH CT        1
              7895 CABER WAY            1
              8020 WALERGA RD           1
              8108 FILIFERA WAY         1
              8721 SPRUCE RIDGE WAY     1
AUBURN        1740 HIGH ST              1
              220 OLD AIRPORT RD        1
              2231 COUNTRY VILLA CT     1
CAMERON PARK  2778 KAWEAH CT            1
              3361 BOW MAR CT           1
              3429 FERNBROOK CT         1
CARMICHAEL    2109 HAMLET PL            1
              4010 ALEX LN              1
              4622 MEYER WAY            1
              5332 SANDSTONE ST         1
              5709 RIVER OAK WAY        1
              5847 DEL CAMPO LN         1
              5907 ELLERSLEE DR         1
                                       ..
SACRAMENTO    8219 GWINHURST CIR        1
              8244 SUNBIRD WAY          1
```

```
                    8299 HALBRITE WAY         1
                    8317 SUNNY CREEK WAY      1
                    8344 FIELDPOPPY CIR       1
                    8345 STAR THISTLE WAY     1
                    8354 SUNRISE WOODS WAY    1
                    8363 LANGTREE WAY         1
                    8421 SUNBLAZE WAY         1
                    8495 DARTFORD DR          1
                    8515 DARTFORD DR          1
                    8593 DERLIN WAY           1
                    8612 WILLOW GROVE WAY     1
                    8728 CRYSTAL RIVER WAY    1
                    8788 LA MARGARITA WAY     1
                    8929 SUTTERS GOLD DR      1
                    9 PASTURE CT              1
                    9012 KIEFER BLVD          1
                    9013 CASALS ST            1
                    9058 MONTOYA ST           1
                    9080 BEDROCK CT           1
                    909 SINGINGWOOD RD        1
                    9186 KINBRACE CT          1
                    923 FULTON AVE            1
                    9257 CALDERA WAY          1
                    9273 PREMIER WAY          1
                    9570 HARVEST ROSE WAY     1
                    9861 CULP WAY             1
WILTON              11215 SHARRMONT CT        1
                    9741 SADDLEBRED CT        1
Length: 496, dtype: int64
```

```
In [7]: pd.crosstab(df.city,df.street)
```

Out[7]:

street	1 KENNELFORD CIR	10 SEA FOAM CT	100 REBECCA WAY	\
city				
ANTELOPE	0	0	0	
AUBURN	0	0	0	
CAMERON PARK	0	0	0	
CARMICHAEL	0	0	0	
CITRUS HEIGHTS	0	0	0	
EL DORADO	0	0	0	
EL DORADO HILLS	0	0	0	
ELK GROVE	0	0	0	
ELVERTA	0	0	0	
FAIR OAKS	0	0	0	
FOLSOM	0	0	1	
GALT	0	0	0	
GOLD RIVER	0	0	0	
GREENWOOD	0	0	0	

LINCOLN	0	0	0
LOOMIS	0	0	0
MATHER	0	0	0
MEADOW VISTA	0	0	0
NORTH HIGHLANDS	0	0	0
ORANGEVALE	0	0	0
PLACERVILLE	0	0	0
POLLOCK PINES	0	0	0
RANCHO CORDOVA	0	0	0
RANCHO MURIETA	0	0	0
RIO LINDA	0	0	0
ROCKLIN	0	0	0
ROSEVILLE	0	0	0
SACRAMENTO	1	1	0
WILTON	0	0	0

street	100 TOURMALINE CIR	10001 WOODCREEK OAKS BLVD Unit 1415	\
city			
ANTELOPE	0	0	
AUBURN	0	0	
CAMERON PARK	0	0	
CARMICHAEL	0	0	
CITRUS HEIGHTS	0	0	
EL DORADO	0	0	
EL DORADO HILLS	0	0	
ELK GROVE	0	0	
ELVERTA	0	0	
FAIR OAKS	0	0	
FOLSOM	0	0	
GALT	0	0	
GOLD RIVER	0	0	
GREENWOOD	0	0	
LINCOLN	0	0	
LOOMIS	0	0	
MATHER	0	0	
MEADOW VISTA	0	0	
NORTH HIGHLANDS	0	0	
ORANGEVALE	0	0	
PLACERVILLE	0	0	
POLLOCK PINES	0	0	
RANCHO CORDOVA	0	0	
RANCHO MURIETA	0	0	
RIO LINDA	0	0	
ROCKLIN	0	0	
ROSEVILLE	0	1	
SACRAMENTO	1	0	
WILTON	0	0	

street	10062 LINCOLN VILLAGE DR	10104 ANNIE ST	10158 CRAWFORD WAY	\
city				
ANTELOPE	0	0	0	
AUBURN	0	0	0	
CAMERON PARK	0	0	0	
CARMICHAEL	0	0	0	
CITRUS HEIGHTS	0	0	0	
EL DORADO	0	0	0	
EL DORADO HILLS	0	0	0	
ELK GROVE	0	1	0	
ELVERTA	0	0	0	
FAIR OAKS	0	0	0	
FOLSOM	0	0	0	
GALT	0	0	0	
GOLD RIVER	0	0	0	
GREENWOOD	0	0	0	
LINCOLN	0	0	0	
LOOMIS	0	0	0	
MATHER	0	0	0	
MEADOW VISTA	0	0	0	
NORTH HIGHLANDS	0	0	0	
ORANGEVALE	0	0	0	
PLACERVILLE	0	0	0	
POLLOCK PINES	0	0	0	
RANCHO CORDOVA	0	0	0	
RANCHO MURIETA	0	0	0	
RIO LINDA	0	0	0	
ROCKLIN	0	0	0	
ROSEVILLE	0	0	0	
SACRAMENTO	1	0	1	
WILTON	0	0	0	

street	1016 CONGRESS AVE	10245 LOS PALOS DR	\
city			
ANTELOPE	0	0	
AUBURN	0	0	
CAMERON PARK	0	0	
CARMICHAEL	0	0	
CITRUS HEIGHTS	0	0	
EL DORADO	0	0	
EL DORADO HILLS	0	0	
ELK GROVE	0	0	
ELVERTA	0	0	
FAIR OAKS	0	0	
FOLSOM	0	0	
GALT	0	0	
GOLD RIVER	0	0	
GREENWOOD	0	0	

LINCOLN	0	0
LOOMIS	0	0
MATHER	0	0
MEADOW VISTA	0	0
NORTH HIGHLANDS	0	0
ORANGEVALE	0	0
PLACERVILLE	0	0
POLLOCK PINES	0	0
RANCHO CORDOVA	0	1
RANCHO MURIETA	0	0
RIO LINDA	0	0
ROCKLIN	0	0
ROSEVILLE	0	0
SACRAMENTO	1	0
WILTON	0	0

street	...	9629 CEDAR OAK WAY \
city	...	
ANTELOPE	...	0
AUBURN	...	0
CAMERON PARK	...	0
CARMICHAEL	...	0
CITRUS HEIGHTS	...	0
EL DORADO	...	0
EL DORADO HILLS	...	0
ELK GROVE	...	1
ELVERTA	...	0
FAIR OAKS	...	0
FOLSOM	...	0
GALT	...	0
GOLD RIVER	...	0
GREENWOOD	...	0
LINCOLN	...	0
LOOMIS	...	0
MATHER	...	0
MEADOW VISTA	...	0
NORTH HIGHLANDS	...	0
ORANGEVALE	...	0
PLACERVILLE	...	0
POLLOCK PINES	...	0
RANCHO CORDOVA	...	0
RANCHO MURIETA	...	0
RIO LINDA	...	0
ROCKLIN	...	0
ROSEVILLE	...	0
SACRAMENTO	...	0
WILTON	...	0

street	9688 NATURE TRAIL WAY	9720 LITTLE HARBOR WAY	\
city			
ANTELOPE	0	0	
AUBURN	0	0	
CAMERON PARK	0	0	
CARMICHAEL	0	0	
CITRUS HEIGHTS	0	0	
EL DORADO	0	0	
EL DORADO HILLS	0	0	
ELK GROVE	1	1	
ELVERTA	0	0	
FAIR OAKS	0	0	
FOLSOM	0	0	
GALT	0	0	
GOLD RIVER	0	0	
GREENWOOD	0	0	
LINCOLN	0	0	
LOOMIS	0	0	
MATHER	0	0	
MEADOW VISTA	0	0	
NORTH HIGHLANDS	0	0	
ORANGEVALE	0	0	
PLACERVILLE	0	0	
POLLOCK PINES	0	0	
RANCHO CORDOVA	0	0	
RANCHO MURIETA	0	0	
RIO LINDA	0	0	
ROCKLIN	0	0	
ROSEVILLE	0	0	
SACRAMENTO	0	0	
WILTON	0	0	

street	9723 TERRAPIN CT	9741 SADDLEBRED CT	9760 LAZULITE CT	\
city				
ANTELOPE	0	0	0	
AUBURN	0	0	0	
CAMERON PARK	0	0	0	
CARMICHAEL	0	0	0	
CITRUS HEIGHTS	0	0	0	
EL DORADO	0	0	0	
EL DORADO HILLS	0	0	0	
ELK GROVE	1	0	1	
ELVERTA	0	0	0	
FAIR OAKS	0	0	0	
FOLSOM	0	0	0	
GALT	0	0	0	
GOLD RIVER	0	0	0	
GREENWOOD	0	0	0	

LINCOLN	0	0	0
LOOMIS	0	0	0
MATHER	0	0	0
MEADOW VISTA	0	0	0
NORTH HIGHLANDS	0	0	0
ORANGEVALE	0	0	0
PLACERVILLE	0	0	0
POLLOCK PINES	0	0	0
RANCHO CORDOVA	0	0	0
RANCHO MURIETA	0	0	0
RIO LINDA	0	0	0
ROCKLIN	0	0	0
ROSEVILLE	0	0	0
SACRAMENTO	0	0	0
WILTON	0	1	0

street	9837 CORTE DORADO CT	9861 CULP WAY	993 MANTON CT	\
city				
ANTELOPE	0	0	0	
AUBURN	0	0	0	
CAMERON PARK	0	0	0	
CARMICHAEL	0	0	0	
CITRUS HEIGHTS	0	0	0	
EL DORADO	0	0	0	
EL DORADO HILLS	0	0	0	
ELK GROVE	1	0	0	
ELVERTA	0	0	0	
FAIR OAKS	0	0	0	
FOLSOM	0	0	0	
GALT	0	0	1	
GOLD RIVER	0	0	0	
GREENWOOD	0	0	0	
LINCOLN	0	0	0	
LOOMIS	0	0	0	
MATHER	0	0	0	
MEADOW VISTA	0	0	0	
NORTH HIGHLANDS	0	0	0	
ORANGEVALE	0	0	0	
PLACERVILLE	0	0	0	
POLLOCK PINES	0	0	0	
RANCHO CORDOVA	0	0	0	
RANCHO MURIETA	0	0	0	
RIO LINDA	0	0	0	
ROCKLIN	0	0	0	
ROSEVILLE	0	0	0	
SACRAMENTO	0	1	0	
WILTON	0	0	0	

```
street           9970 STATE HIGHWAY 193
city
ANTELOPE                              0
AUBURN                                0
CAMERON PARK                          0
CARMICHAEL                            0
CITRUS HEIGHTS                        0
EL DORADO                             0
EL DORADO HILLS                       0
ELK GROVE                             0
ELVERTA                               0
FAIR OAKS                             0
FOLSOM                                0
GALT                                  0
GOLD RIVER                            0
GREENWOOD                             0
LINCOLN                               0
LOOMIS                                0
MATHER                                0
MEADOW VISTA                          0
NORTH HIGHLANDS                       0
ORANGEVALE                            0
PLACERVILLE                           1
POLLOCK PINES                         0
RANCHO CORDOVA                        0
RANCHO MURIETA                        0
RIO LINDA                             0
ROCKLIN                               0
ROSEVILLE                             0
SACRAMENTO                            0
WILTON                                0

[29 rows x 496 columns]
```

In [4]:
```
df["city"].value_counts()
```

Out[4]:
```
SACRAMENTO         248
ELK GROVE           66
CITRUS HEIGHTS      19
ANTELOPE            17
ROSEVILLE           17
RANCHO CORDOVA      15
GALT                14
NORTH HIGHLANDS     13
FOLSOM              10
CARMICHAEL          10
RIO LINDA           10
EL DORADO HILLS      8
```

```
LINCOLN              7
PLACERVILLE          5
ROCKLIN              5
ORANGEVALE           5
FAIR OAKS            5
ELVERTA              4
GOLD RIVER           3
CAMERON PARK         3
AUBURN               3
RANCHO MURIETA       3
POLLOCK PINES        2
WILTON               2
GREENWOOD            1
LOOMIS               1
MATHER               1
MEADOW VISTA         1
EL DORADO            1
Name: city, dtype: int64
```

```
In [66]: df["type"].value_counts()
```

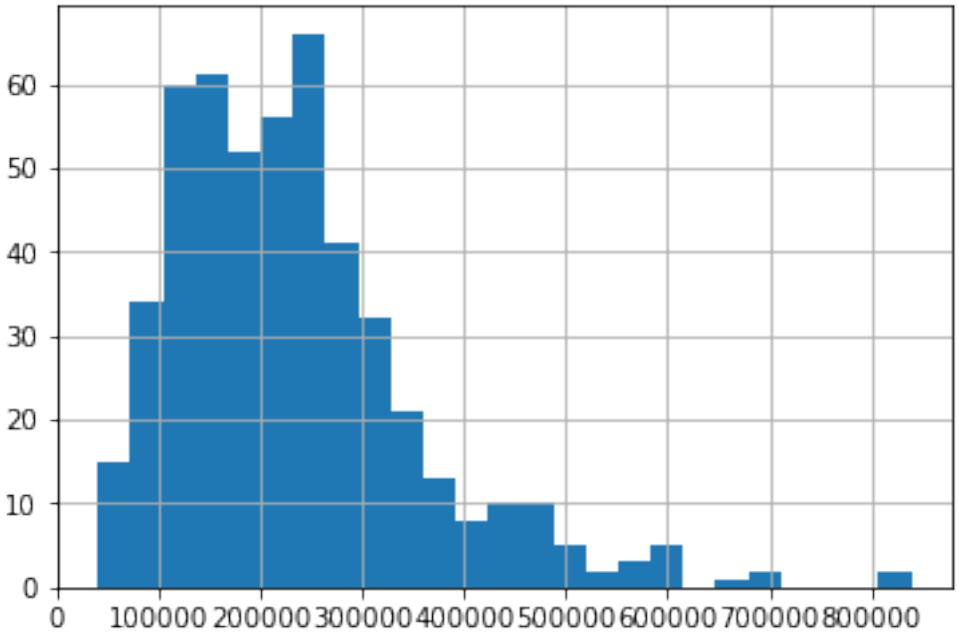
```
Out[66]: Residential     453
         Condo            37
         Multi-Family      9
         Name: type, dtype: int64
```

```
In [8]: df.loc[(df["city"]=="SACRAMENTO") &(df["type"]=="Condo")& (df['price']<=150000),['city
```

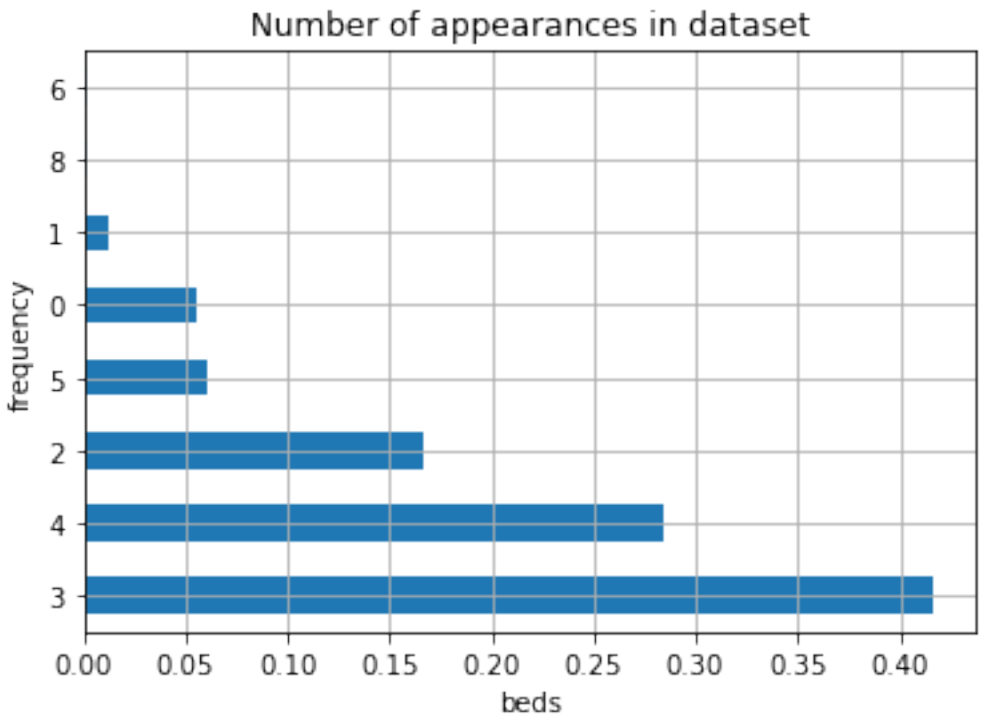
```
Out[8]:            city   type   price
        5    SACRAMENTO  Condo   89921
        15   SACRAMENTO  Condo  110700
        39   SACRAMENTO  Condo  150000
        335  SACRAMENTO  Condo   40000
        336  SACRAMENTO  Condo   48000
        343  SACRAMENTO  Condo   77000
        344  SACRAMENTO  Condo   82732
        349  SACRAMENTO  Condo   90000
        354  SACRAMENTO  Condo  100000
        368  SACRAMENTO  Condo  120000
        377  SACRAMENTO  Condo  126960
```

```
In [9]: #Creating a new function:
        def num_missing(x):
              return sum(x.isnull())

        #Applying per column:
        print("Missing values per column:")
        print (df.apply(num_missing, axis=0)) #axis=0 defines that function is to be applied o
```

```
        print("\nMissing values per row:")
        print(df.apply(num_missing, axis=1).head()) #axis=1 defines that function is to be app

Missing values per column:
street       0
city         0
zip          0
state        0
beds         0
baths        0
sq__ft       0
type         0
sale_date    0
price        0
latitude     0
longitude    0
dtype: int64

Missing values per row:
0    0
1    0
2    0
3    0
4    0
dtype: int64


In [10]:  df.apply(lambda x: sum(x.isnull()),axis=0) #using lambda

Out[10]: street       0
         city         0
         zip          0
         state        0
         beds         0
         baths        0
         sq__ft       0
         type         0
         sale_date    0
         price        0
         latitude     0
         longitude    0
         dtype: int64

In [11]: %matplotlib inline
         import matplotlib.pyplot as plt
         df["price"].hist(bins=25)#histogram of the first 25 data of the price

Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x118c5ba20>
```

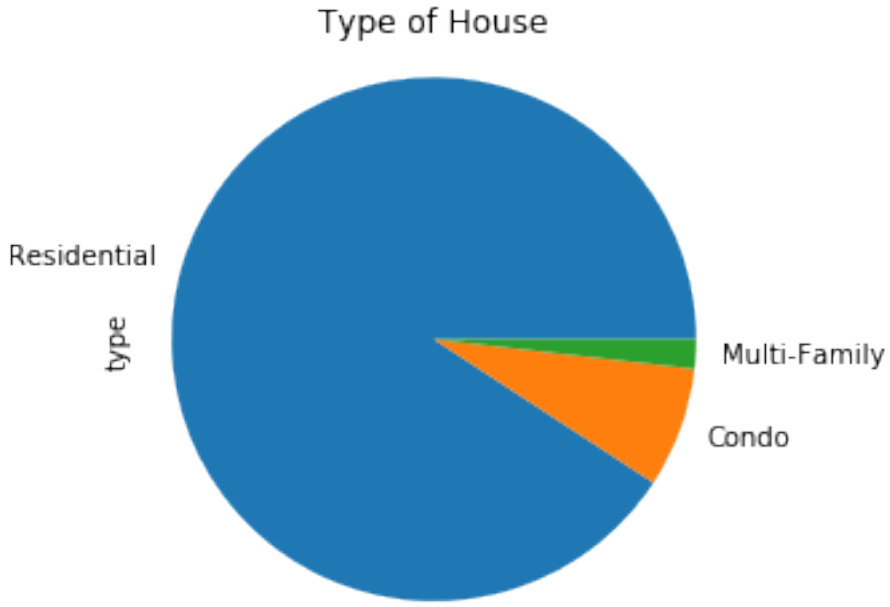



```
In [42]: df.beds.value_counts(25).plot(kind='barh',grid=True)
         plt.title('Number of appearances in dataset')
         plt.xlabel('beds')
         plt.ylabel('frequency') #histogram of city.Incorrect graph.

Out[42]: <matplotlib.text.Text at 0x11dc67a90>
```



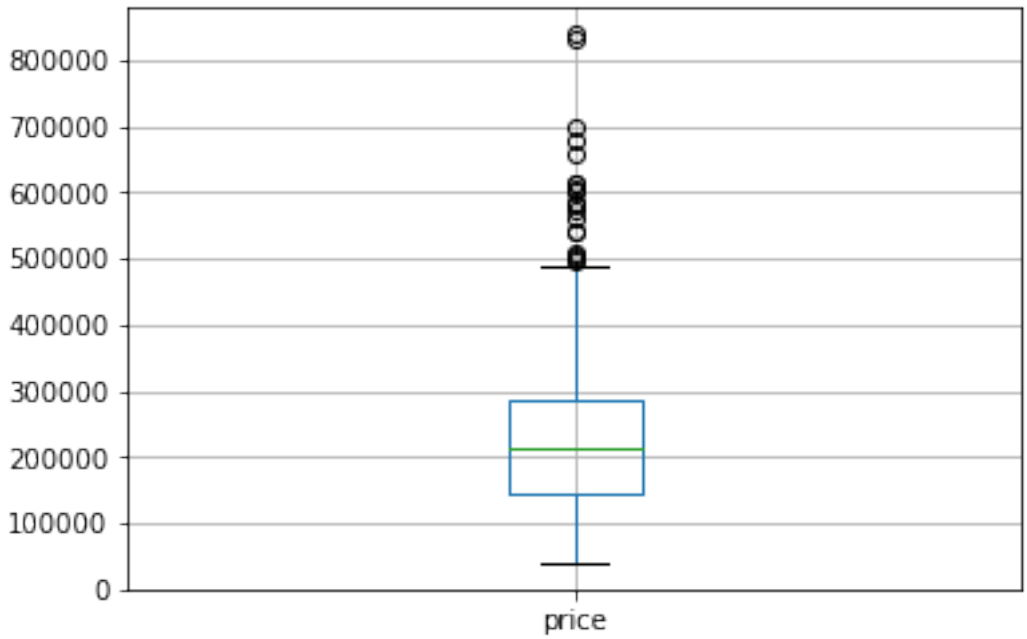

```
In [16]: df.type.value_counts().plot(kind='pie')
         plt.axis('scaled')
         plt.title('Type of House')#pie chart version of the type of house
```

```
Out[16]: <matplotlib.text.Text at 0x11c5984e0>
```



```
In [18]: df.boxplot(column='price')
```

```
Out[18]: <matplotlib.axes._subplots.AxesSubplot at 0x11c699b38>
```



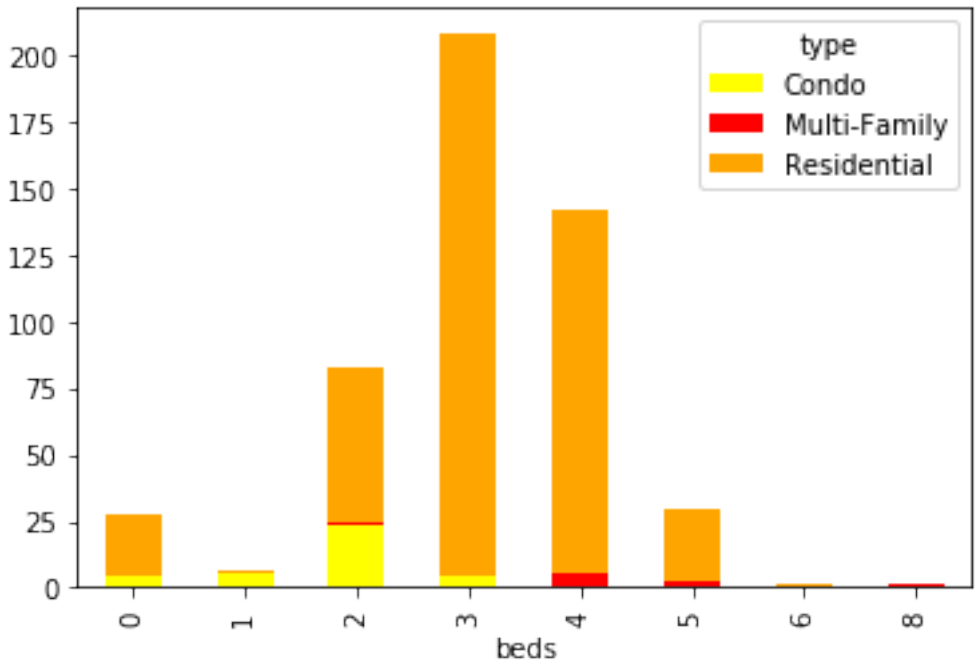

```
In [33]: #pivot table
         temp1 = df['beds'].value_counts(ascending=True)
         print('Frequency Table for Bedrooms:' )
         print(temp1)
         temp2 = df.pivot_table(values='type',index=['beds'],aggfunc=lambda x: x.map({'Resident
         print(temp2) #not sure about temp2 did the same as intructed on the website.
```

```
Frequency Table for Bedrooms:
6      1
8      1
1      6
0     28
5     30
2     83
4    142
3    208
Name: beds, dtype: int64
          type
beds
```

```
0     0.857143
1     0.166667
2     0.707317
3     0.980769
4     1.000000
5     1.000000
6     1.000000
8          NaN
```

```
In [80]: temp3=pd.crosstab(df['beds'],df['type'])
         temp3.plot(kind='bar',stacked=True, color=['Yellow','red','orange'],grid=False)#stack
```

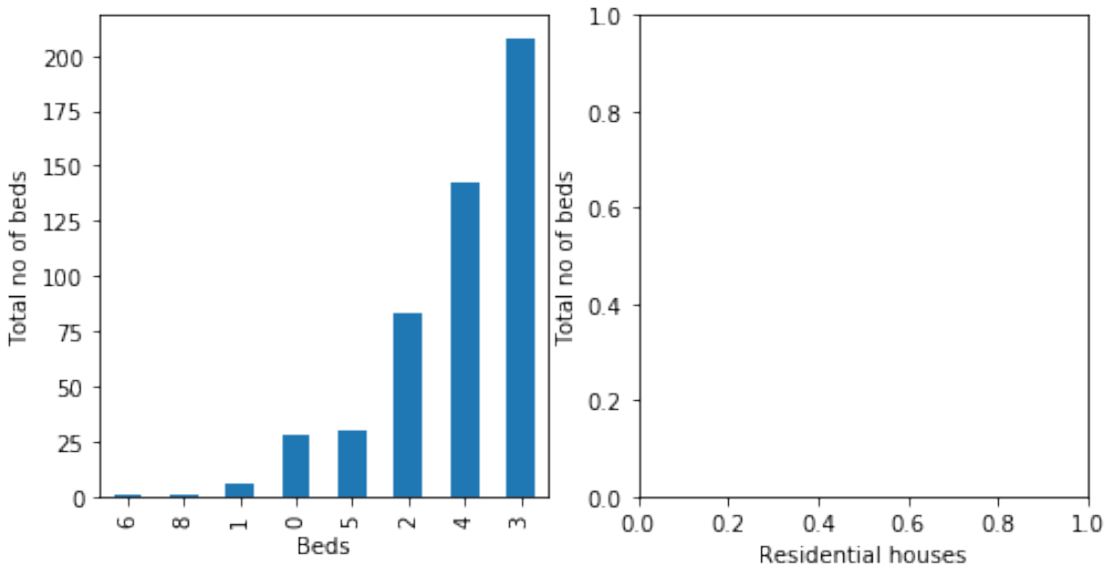
```
Out[80]: <matplotlib.axes._subplots.AxesSubplot at 0x11dff0b00>
```

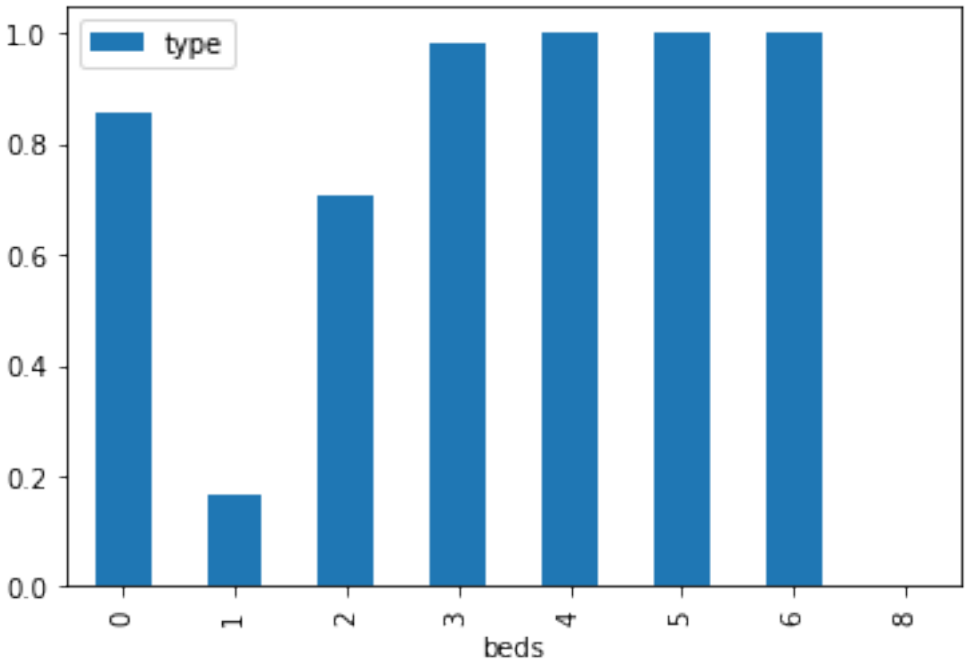



```
In [47]: import matplotlib.pyplot as plt
         fig = plt.figure(figsize=(8,4))
         ax1 = fig.add_subplot(121)
         ax1.set_xlabel('Beds')
         ax1.set_ylabel('Total no of beds')
         ax1.set_title("")
         temp1.plot(kind='bar')

         ax2 = fig.add_subplot(122)
         temp2.plot(kind = 'bar')
```

```
        ax2.set_xlabel('Residential houses')
        ax2.set_ylabel('Total no of beds')
        ax2.set_title("")
```

```
Out[47]: <matplotlib.text.Text at 0x11ea254e0>
```





```
In [52]: pd.crosstab(df["beds"],df["type"],margins=False) #crosstab
```

Out[52]:

beds	Condo	Multi-Family	Residential
0	4	0	24
1	5	0	1
2	24	1	58
3	4	0	204
4	0	5	137
5	0	2	28
6	0	0	1
8	0	1	0

```
In [53]: pd.crosstab(df["beds"],df["type"],margins=True) #crosstab
```

Out[53]:

beds	Condo	Multi-Family	Residential	All
0	4	0	24	28
1	5	0	1	6
2	24	1	58	83
3	4	0	204	208
4	0	5	137	142
5	0	2	28	30
6	0	0	1	1
8	0	1	0	1
All	37	9	453	499

```
In [58]: prop_rates = pd.DataFrame([1000, 5000, 12000], index=['Residential','COndo','Multi-Fa
         prop_rates
```

Out[58]:

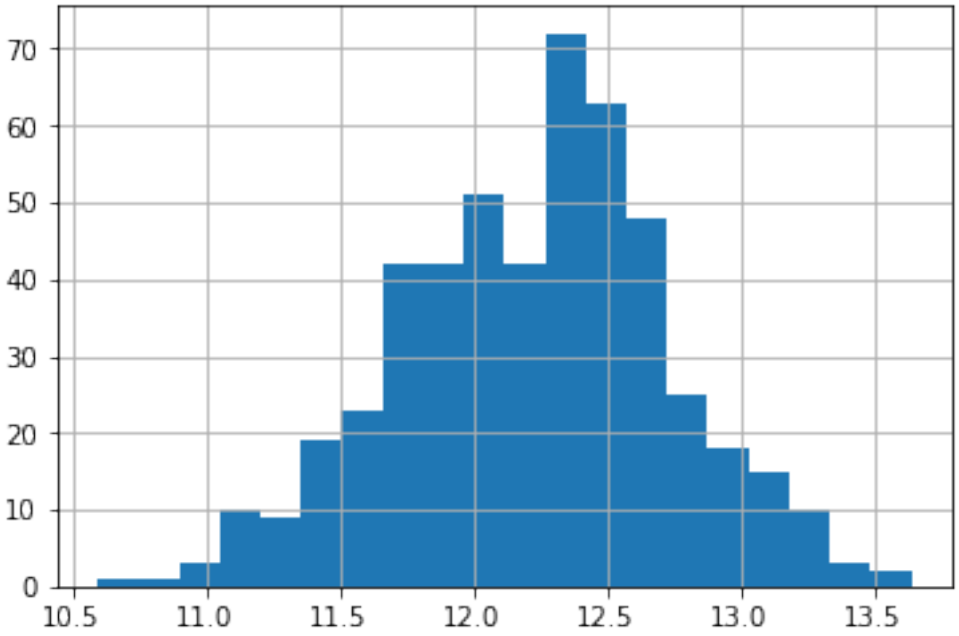
	rates
Residential	1000
COndo	5000
Multi-Family	12000

```
In [62]: data_sorted = df.sort_values(['beds','baths'], ascending=False)
         data_sorted[['beds','baths']].head(10)
         #sorted the datas
```

Out[62]:

	beds	baths
108	8	4
157	6	5
332	5	5
120	5	4
151	5	4
258	5	4
303	5	4
307	5	4
451	5	4
78	5	3

```
In [68]: df['price_log'] = np.log(df['price'])
         df['price_log'].hist(bins=20)#used log for extreme numbers

Out[68]: <matplotlib.axes._subplots.AxesSubplot at 0x11e5fff60>
```




```
In [78]: df.dtypes

Out[78]: street        object
         city          object
         zip            int64
         state         object
         beds           int64
         baths          int64
         sq__ft         int64
         type          object
         sale_date     object
         price          int64
         latitude     float64
         longitude    float64
         price_log    float64
         dtype: object

In [89]: import numpy as np

         import matplotlib as plt
```

```
from sklearn import datasets

from sklearn import metrics

from sklearn.linear_model import LogisticRegression #importing scikit
```

In [111]:
```
from sklearn.preprocessing import LabelEncoder
var_mod = ['street','state','city','type','sale_date']
le = LabelEncoder()
for i in var_mod:
    df[i] = le.fit_transform(df[i])
df.dtypes
```

Out[111]:
```
street              int64
city                int64
zip                 int64
state               int64
beds                int64
baths               int64
sq__ft              int64
type                int64
sale_date           int64
price               int64
latitude          float64
longitude         float64
price_log         float64
Totalbed&baths      int64
TotalIncome         int64
TotalIncome_log   float64
dtype: object
```

In [123]:
```
#i tried decision tree, logistic regression but it didn't work
```

In [ ]:

In [ ]: