

Exploratory Data Analysis OF Lagos House Prices

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Objective

To determine the best place to live in lagos based on the following parameters

- Security
- Budget
- Other Utilities
- House Density
- How Landlords can Increase their Prices

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Introduction

Lagos is state in the south western part of Nigeria, that boasts of the highest population density of any city in Africa. Here the the cost of renting houses are quite high, so I decided analyze the cost of owning a house in the city. We are hoping to look into the pricing patterns of lagos houses draw observation and render conclusion. This dataset was provided by quantum analytics

```
In [2]: # Importing Libraries
import numpy as np
import pandas as pd
```

```
import matplotlib.pyplot as plt
import seaborn as sns
```

Data Wrangling

```
In [4]: # read file into pandas dataframe
lagos_houses = pd.read_csv(r'C:\Users\Mhizfair\Desktop\Quantum analytics\Python\lagos_house_prices_raw.csv')
```

```
In [5]: lagos_houses.head()
```

```
Out[5]:
```

	location	bed	bath	toilet	price	Property_Type	Parking_Space	Security	Electricity	Furnished	Security_Doors	CCTV	Pool	Gym	BQ
0	yaba	1	1	2	700000.0	Mini flat	0	0	0	0	0	0	0	0	0
1	yaba	1	1	2	700000.0	Mini flat	0	0	0	0	0	0	0	0	0
2	yaba	1	1	2	650000.0	Mini flat	0	0	0	0	0	0	0	0	0
3	yaba	1	1	1	450000.0	Mini flat	0	0	0	0	0	0	0	0	0
4	yaba	3	3	4	800000.0	Detached duplex	0	1	0	0	0	0	0	0	0

```
In [6]: lagos_houses.shape
```

```
Out[6]: (5336, 15)
```

```
In [7]: lagos_houses.dtypes
```

```
Out[7]: location      object
        bed          int64
        bath         int64
        toilet       int64
        price        float64
        Property_Type object
        Parking_Space int64
        Security      int64
        Electricity   int64
        Furnished     int64
        Security_Doors int64
        CCTV        int64
        Pool          int64
        Gym           int64
        BQ            int64
        dtype: object
```

```
In [8]: lagos_houses.columns
```

```
Out[8]: Index(['location', 'bed', 'bath', 'toilet', 'price', 'Property_Type',
              'Parking_Space', 'Security', 'Electricity', 'Furnished',
              'Security_Doors', 'CCTV', 'Pool', 'Gym', 'BQ'],
              dtype='object')
```

```
In [9]: # Checking if data is duplicated
        sum(lagos_houses.duplicated())
```

```
Out[9]: 3434
```

```
In [21]: # Viewing duplicate File
        lagos_houses[lagos_houses.duplicated()]
```

Out[21]:

	location	bed	bath	toilet	price	Property_Type	Parking_Space	Security	Electricity	Furnished	Security_Doors	CCTV	Pool	Gym	BQ
1	yaba	1	1	2	700000.0	Mini flat	0	0	0	0	0	0	0	0	0
8	yaba	1	1	2	700000.0	Mini flat	0	0	0	0	0	0	0	0	0
9	yaba	1	1	2	700000.0	Mini flat	0	0	0	0	0	0	0	0	0
22	yaba	1	1	1	350000.0	Self contain	0	0	0	0	0	0	0	0	0
23	yaba	1	1	2	700000.0	Mini flat	0	0	0	0	0	0	0	0	0
...
5327	ajah	1	1	1	600000.0	Mini flat	0	0	0	0	0	0	0	0	0
5329	ajah	1	1	2	500000.0	Mini flat	1	0	0	0	0	0	0	0	0
5330	ajah	1	1	2	550000.0	Mini flat	0	0	0	0	0	0	0	0	0
5333	ajah	4	4	5	1700000.0	Semi detached duplex	1	0	0	0	0	0	0	0	0
5334	ajah	1	1	2	500000.0	Mini flat	0	0	0	0	0	0	0	0	0

3434 rows × 15 columns

In [13]:

```
# View Locations In Lagos
lagos_houses.location.unique()
```

Out[13]:

```
array(['yaba', 'surulere', 'gbagada', 'lekki phase 1', 'ikorodu', 'ikeja',
      'ajah'], dtype=object)
```

In [18]:

```
# Price Of Most Expensive House
lagos_houses.price.max()
```

Out[18]:

```
2450000.0
```

In [22]:

```
lagos_houses.price.min()
```

Out[22]:

```
150.0
```

In [25]:

```
columns = ['Parking_Space', 'Security', 'Furnished', 'Electricity', 'Security_Doors', 'CCTV', 'Pool', 'Gym', 'BQ' ]
for col in columns:
```

```
lagos_houses[col].replace({0: 'no', 1: 'yes'}, inplace = True)
```

```
In [27]: lagos_houses.head()
```

```
Out[27]:
```

	location	bed	bath	toilet	price	Property_Type	Parking_Space	Security	Electricity	Furnished	Security_Doors	CCTV	Pool	Gym	BQ
0	yaba	1	1	2	700000.0	Mini flat	no	no	no	no	no	no	no	no	no
1	yaba	1	1	2	700000.0	Mini flat	no	no	no	no	no	no	no	no	no
2	yaba	1	1	2	650000.0	Mini flat	no	no	no	no	no	no	no	no	no
3	yaba	1	1	1	450000.0	Mini flat	no	no	no	no	no	no	no	no	no
4	yaba	3	3	4	800000.0	Detached duplex	no	yes	no	no	no	no	no	no	no

```
In [26]: lagos_houses.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5336 entries, 0 to 5335
Data columns (total 15 columns):
 #   Column                Non-Null Count  Dtype  
---  -
 0   location              5336 non-null  object  
 1   bed                   5336 non-null  int64   
 2   bath                  5336 non-null  int64   
 3   toilet                5336 non-null  int64   
 4   price                 5336 non-null  float64  
 5   Property_Type         5336 non-null  object  
 6   Parking_Space         5336 non-null  object  
 7   Security              5336 non-null  object  
 8   Electricity           5336 non-null  object  
 9   Furnished            5336 non-null  object  
10  Security_Doors        5336 non-null  object  
11  CCTV                5336 non-null  object  
12  Pool                  5336 non-null  object  
13  Gym                   5336 non-null  object  
14  BQ                    5336 non-null  object  
dtypes: float64(1), int64(3), object(11)
memory usage: 625.4+ KB
```

What is the structure of your dataset? There were 5336 records of data and 15 features (location,bed,bath,toilet,price, Property_Type,Parking_Space,Security,Electricity,Furnished,Security_Doors,CCTV,Pool,Gym,BQ). The variables can be grouped as numeric and

categorical.

(1=Yes) —> (0=No) What is/are the main feature(s) of interest in your dataset? I'm interested in figuring out what features are best for predicting the prices of rent for Houses in lagos.

What features in the dataset do you think will help support your investigation into your feature(s) of interest? I expect that blood pressure will have the strongest effect on cvd: the higher the blood pressure, the greater risk of cvd. I also think that age, weight, height, cholesterol levels would have contributing effects.

Univariate Analysis:

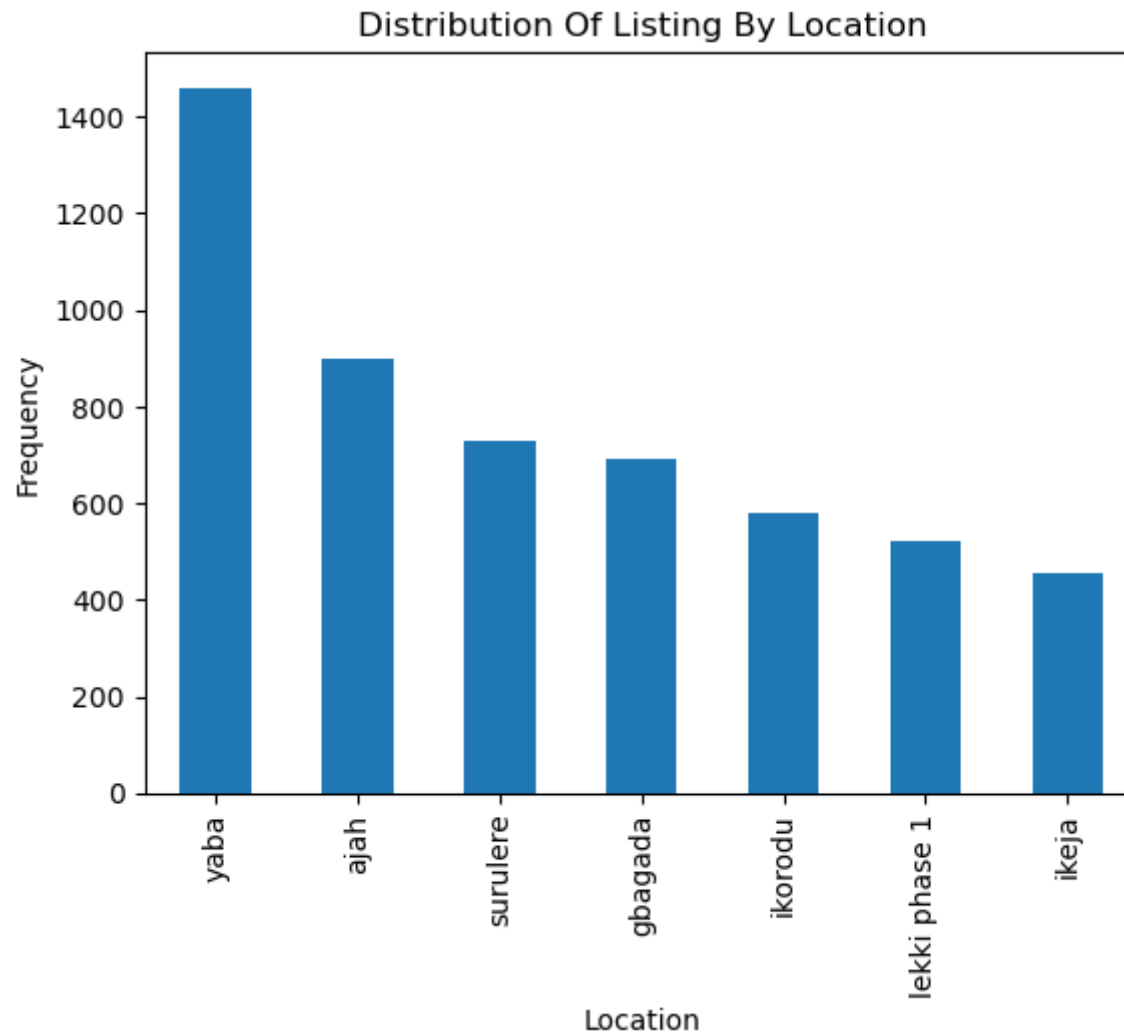
This helps to get an overview of each variable in the data.

```
In [33]: # Lets see how each location listed their houses
count_listing = lagos_houses['location'].value_counts()
count_listing
```

```
Out[33]: yaba          1460
ajah          900
surulere      728
gbagada       692
ikorodu       578
lekki phase 1  521
ikeja         457
Name: location, dtype: int64
```

```
In [38]: count_listing.plot.bar(title= 'Distribution Of Listing By Location', ylabel= 'Frequency', xlabel= 'Location')

plt.show()
```



Observation

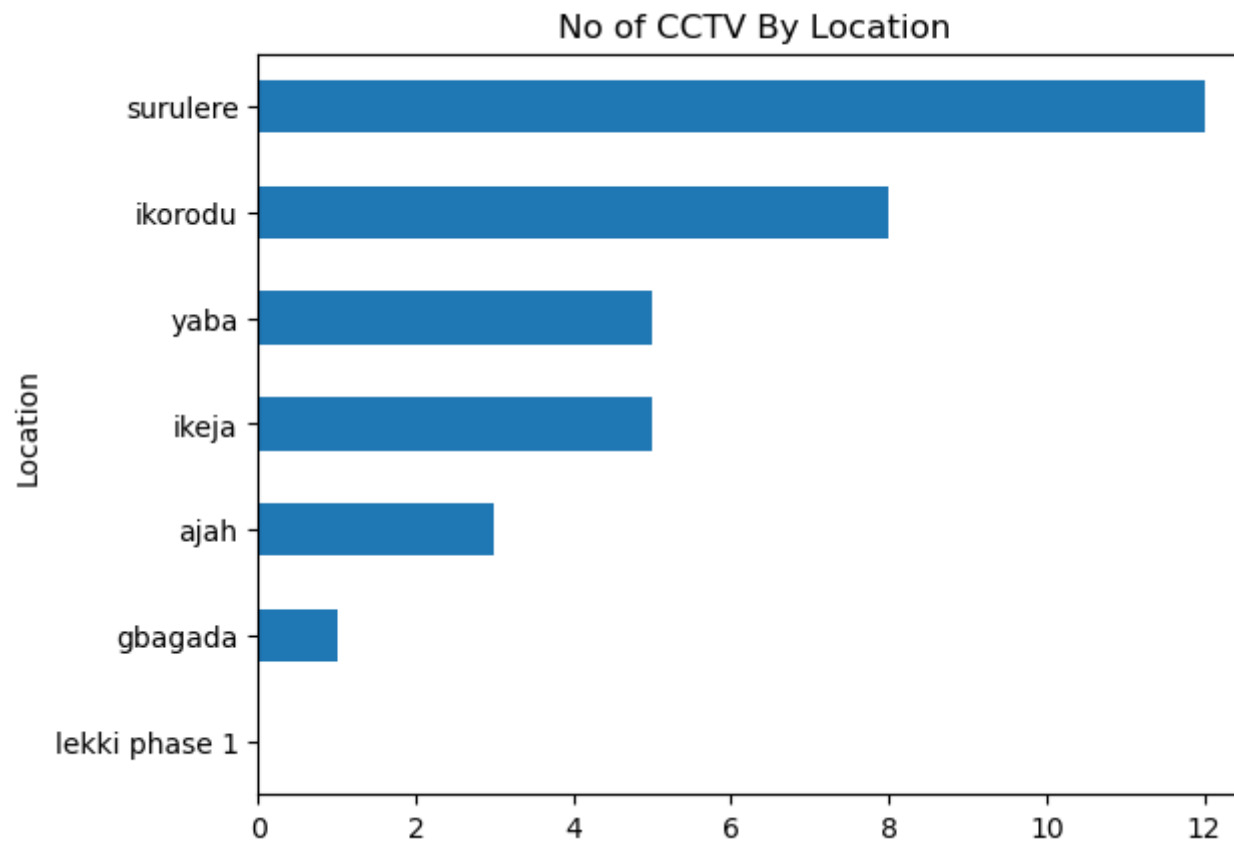
From the above barchart we can see that yaba had the highest amount of listing. Later we would investigate if the prices of the houses in yaba is responsible for the high amount of listing there also to note is that Ikeja had the least amount of listing despite being the state capital

```
In [60]: # Can we deduce the Location that has CCTV?
ctv= lagos_houses.groupby('location')['CCTV'].apply(lambda x: (x== 'yes').sum()).sort_values()
```

```
cctv
```

```
Out[60]: location  
lekki phase 1    0  
gbagada          1  
ajah             3  
ikeja            5  
yaba             5  
ikorodu          8  
surulere         12  
Name: CCTV, dtype: int64
```

```
In [61]: cctv.plot.barh(title= 'No of CCTV By Location', ylabel= 'Frequency', xlabel= 'Location')  
plt.show()
```

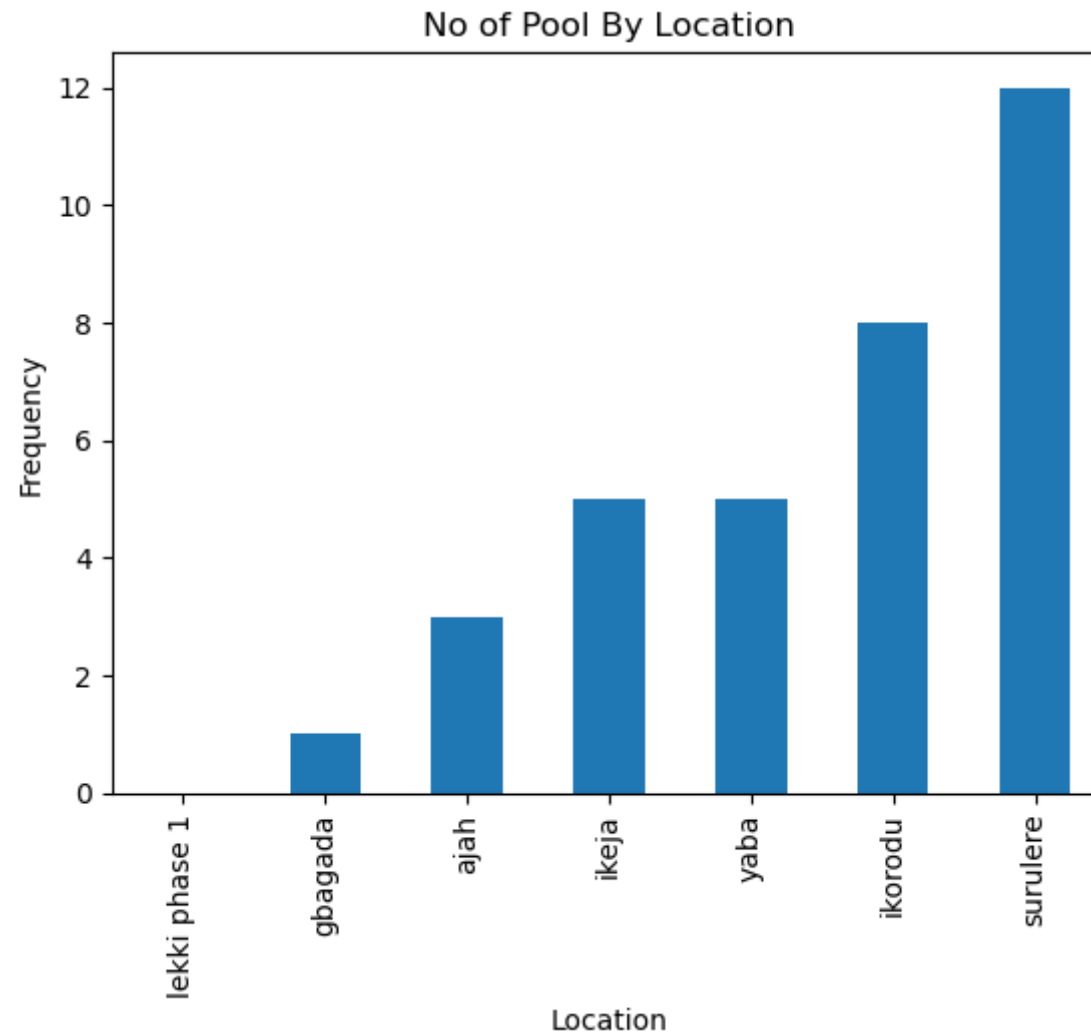



```
In [59]: # No. Of Locations with Pools
pool = lagos_houses.groupby('location')['Pool'].apply(lambda x: (x == 'yes').sum())
pool
```

```
Out[59]: location
ajah      1
gbagada   7
ikeja     5
ikorodu   0
lekki phase 1  15
surulere   3
yaba      1
Name: Pool, dtype: int64
```

```
In [63]: cctv.plot.bar(title= 'No of Pool By Location', ylabel= 'Frequency', xlabel= 'Location')

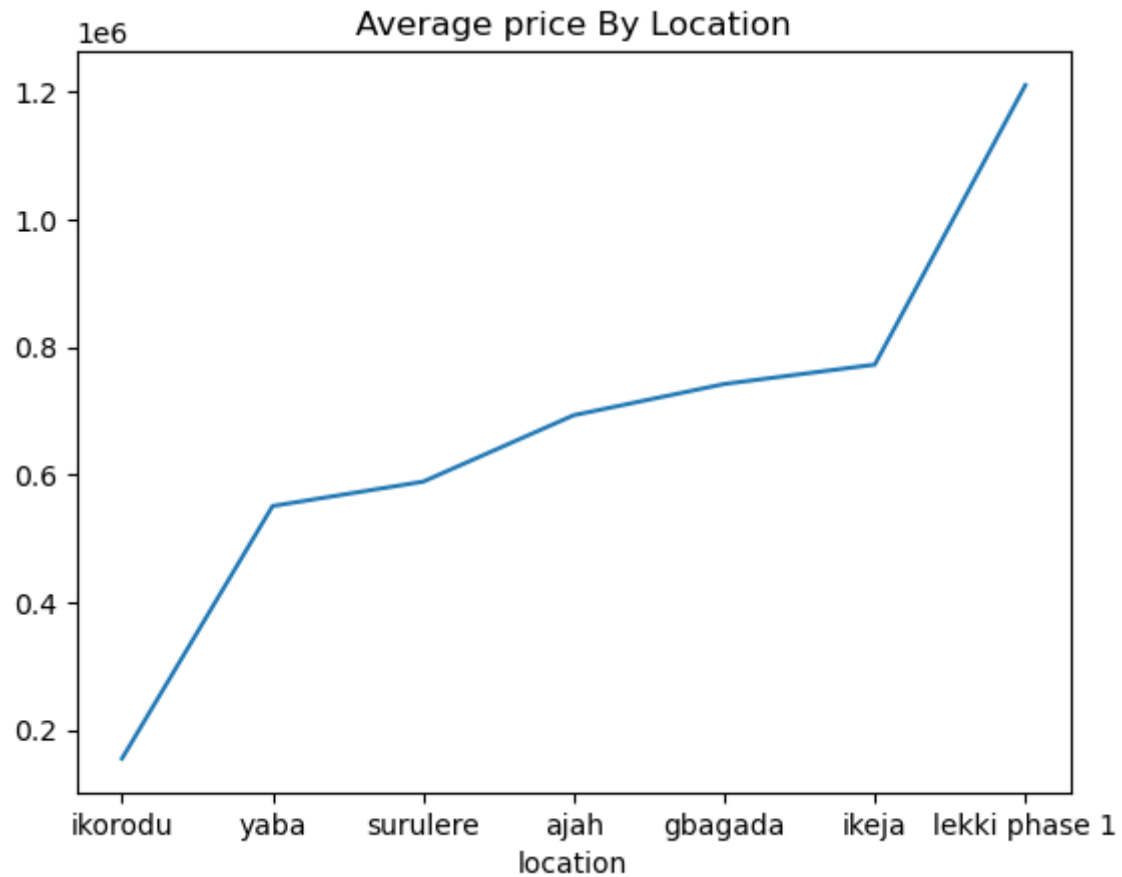
plt.show()
```



```
In [68]: # Average Price of Houses by Location
Avg_price= lagos_houses['price'].groupby(lagos_houses.location).mean().astype('int').sort_values()
Avg_price
```

```
Out[68]: location
ikorodu      155095
yaba         550986
surulere     589189
ajah         693335
gbagada      742290
ikeja        772700
lekki phase 1 1211013
Name: price, dtype: int32
```

```
In [82]: Avg_price.plot(kind='line', x='location', y='price', title='Average price By Location')
plt.show()
```



Observation

So far we can deduce that Yaba has the highest amount of listing, Surulere has the highest amount of houses that has CCTV in them as much as 12 listings from Surulere boast of CCTV its rather unclear why Lekki phase 1 has no CCTV or pool despite being the highest the most expensive when we look at the average price by location. I think so far it is safe to say that Swimming pool and CCTV has little or no impact on the prices of houses as Surulere who has the highest amount in those have just a little below average when we look at the average of house prices based on location. As we investigate further lets see if we can find relationships that we could link to the bizzare high cost of prices in Lekki Phase 1

Bivariate Analysis

-Lets investige to see if we can link anything in the data to the high cost of houses in lekki phase 1.

```
In [84]: # Before we go deep into it Lets Look at the type of Property that has the highest amount of average price
Prop_type= lagos_houses['price'].groupby(lagos_houses.Property_Type).mean().astype('int').sort_values()
Prop_type
```

```
Out[84]: Property_Type
Self contain          447286
Mini flat             590483
Flat / apartment      977908
Blocks of flats       1108333
Penthouse flat        1160714
Semi detached bungalow 1200000
Detached bungalow     1313888
Terraced duplex       1632894
Detached duplex       1695396
Semi detached duplex  1747878
Massionette house     2200000
Name: price, dtype: int32
```

There is no surprise here as the average prices of each type of property tend to increase as the property type grew bigger. Here the most expensive type of proprty was unsurprisingly the Massionette type of houses while the least expensive was the Self contain apartments

```
In [124... ### Now Lets see if we can see the type of properties Listed per Location
counts=lagos_houses.groupby(['location', 'Property_Type']).size().reset_index(name= 'prop')
table= counts.pivot(index= 'location', columns= 'Property_Type', values= 'prop')
table=table.fillna(0)
table.head(7)
```

Out[124]:

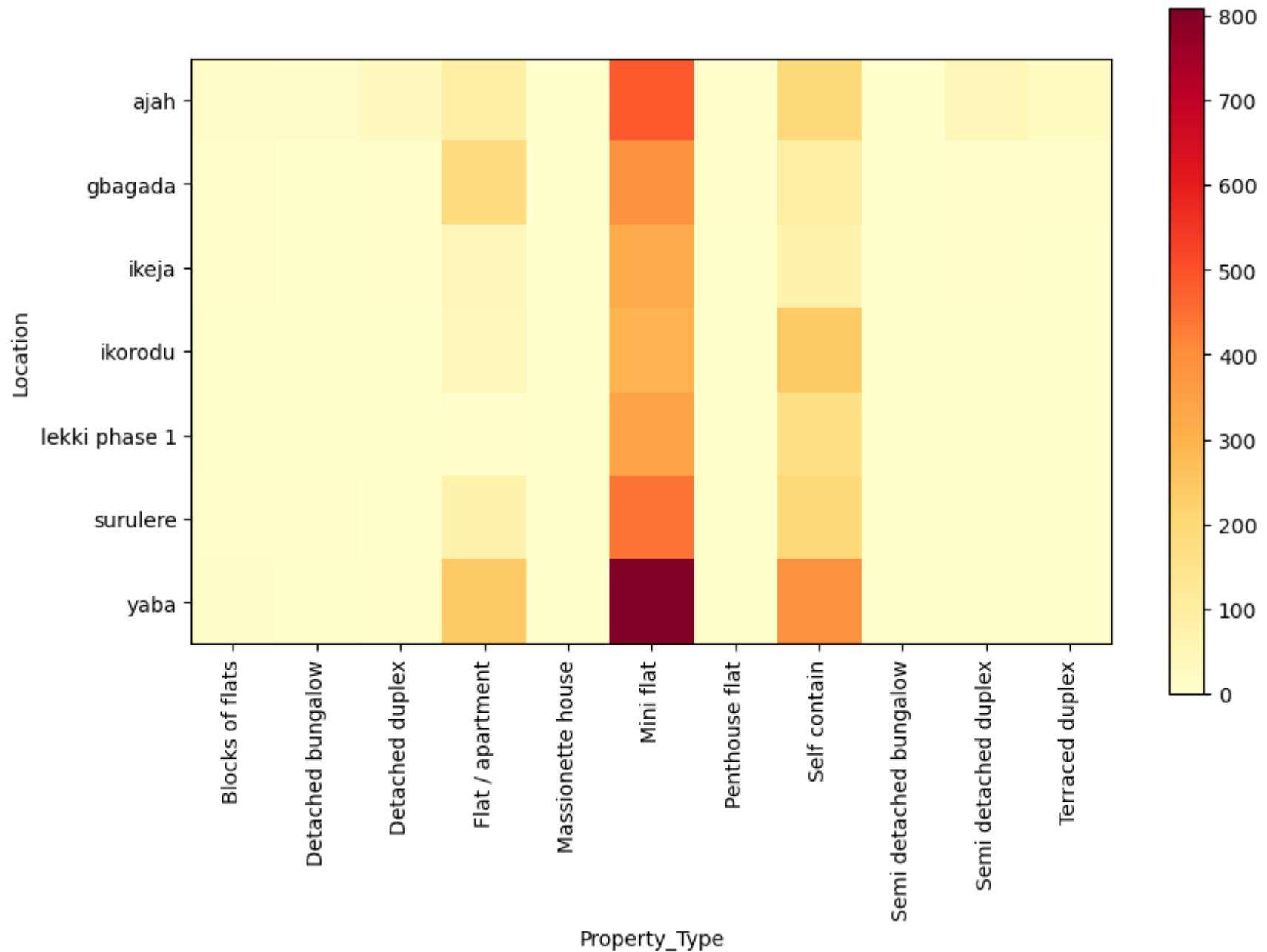
Property_Type	Blocks of flats	Detached bungalow	Detached duplex	Flat / apartment	Massionette house	Mini flat	Penthouse flat	Self contain	Semi detached bungalow	Semi detached duplex	Terraced duplex
location											
ajah	15.0	10.0	29.0	89.0	0.0	483.0	5.0	191.0	3.0	47.0	28.0
gbagada	6.0	1.0	4.0	186.0	0.0	387.0	5.0	90.0	0.0	5.0	8.0
ikeja	6.0	1.0	8.0	49.0	0.0	316.0	0.0	70.0	0.0	6.0	1.0
ikorodu	0.0	1.0	0.0	40.0	0.0	296.0	0.0	240.0	0.0	1.0	0.0
lekki phase 1	0.0	1.0	2.0	9.0	1.0	336.0	1.0	167.0	0.0	3.0	1.0
surulere	4.0	4.0	3.0	73.0	0.0	444.0	1.0	196.0	0.0	3.0	0.0
yaba	11.0	0.0	7.0	240.0	0.0	808.0	2.0	391.0	0.0	1.0	0.0

In [139...

```

fig, ax= plt.subplots(figsize=(10, 6))
plt.rcParams.update({'font.size': 10})
im= ax.imshow(table, cmap='YlOrRd', interpolation='nearest')
ax.set_xticks(range(len(table.columns)))
ax.set_yticks(range(len(table.index)))
ax.set_xticklabels(table.columns, rotation=90)
ax.set_yticklabels(table.index)
ax.set_xlabel('Property_Type')
ax.set_ylabel('Location')
fig.colorbar(im)
plt.show()

```



If you look at the table closely you'd discover that we have an answer to why Lekki Phase 1 has a very high average price compared to others despite not having swimming pools and CCTV cameras like Surulere. This is because Lekki Phase 1 is the only location that boasts of a mansion and

from the property type analysis we did we learnt that that property is outrageously expensive. Ajah seems to be the location that has a perfect distribution for different property Types. Lets dig deeper and see if security has any effect on price

In [147...

```
S= lagos_houses.groupby('location')['Security'].apply(lambda x: (x== 'yes').sum())
S_Doors=pool= lagos_houses.groupby('location')['Security_Doors'].apply(lambda x: (x== 'yes').sum())
print(S)
print(S_Doors)
```

```
location
ajah          83
gbagada       43
ikeja         37
ikorodu       93
lekki phase 1  45
surulere      39
yaba         218
Name: Security, dtype: int64
location
ajah          5
gbagada       0
ikeja         0
ikorodu       0
lekki phase 1  0
surulere      0
yaba          0
Name: Security_Doors, dtype: int64
```

Conclusion

From the above exploratory analysis I was able to deduce the following to answer the Analysis Objective

- Yaba is the mos densely concentrated amount of listing with over 700+ house listing
- Lekki Phase 1 is the most expensive neighborhood acoording to the dataset as it has the highest amount of average price(though this is largely due to the fact that it is the only location that has a moansion and mansion is the most expensive kind of property.
- Mini Flat is the most comon type of property in lagos and every single location has atleast 200 mini flats while Mansion is the least popular kind of propert with just one mansion in the whole of lagos.
- Ajah is the most security conscious neighborhood with a good number of security outfits and the land lords here are very conscious of installing security doors as landlords in other areas did not install any security door.

- CCTV, Swimming pools and other utilities has little or no effect on the prices of property.

Recommendations

After making the above conclusions I have the following recommendations to make to the landlords

- Landlords should decongest Ajah as the amount of listing in Ajah is too much and excess demand in a location seems to reduce the price of houses in that area
- While building Landlords could concentrate more on security doors rather than utilities like swimming pools and CCTV as the later has little or no impact on the price of property.
- Priorities should be put in building other types of propertyies such as pent houses because houses like mini flats are already over saturated
- Landlords in Lekki Phase one should prioritize installing CCTV cameras and swimming pools into their properties as this would help the price appreciation of their properties

In []:

In []:

In []: