Introduction:

In this report, I will analyze a dataset containing real estate data including property details, pricing, bedrooms, bathrooms, and more. The dataset is loaded using the pandas library and consists of various features such as location, price, property type, area, and seller type. Our objective is to uncover insights and answer specific questions regarding the properties in the dataset.

```
In [1]: #import Libraries:
    import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    import warnings
    warnings.filterwarnings('ignore')
```

About Sales Data

Data Loading and Initial Exploration:

```
In [2]: # Read the dataset:
   data = pd.read_excel(r"data test Sales.xlsx")
```

In [3]: data

10

19

Cairo

Out[3]: Property_ID Location Property_Type Listing_Type Price **Bedrooms Bathrooms** Seller_Type Area 0 Cairo Villa Sale 5000000 4 3 250 Individual 2 2 1 Giza Apartment Sale 1500000 100 Developer 2 5 3000000 3 2 Giza Villa 200 Individual Sale 1200000 2 3 6 Alexandria Apartment Sale 1 90 Developer 7 5 4 Villa 4500000 300 Cairo Sale Individual 4 5 10 Apartment 2000000 3 2 150 Developer Cairo Sale 6 11 Villa 3800000 4 3 230 Individual Giza Sale 7 1600000 3 2 130 Developer 14 Giza Apartment Sale 8 5500000 5 320 Alexandria Villa Sale 4 Individual 9 18 Alexandria 1800000 2 1 110 Individual Apartment Sale

```
In [4]: # Display the first 5 rows of the dataset:
    data.head()
```

Villa

Sale 4200000

4

3

280

Developer

Out[4]:		Property_ID	Location	Property_Type	Listing_Type	Price	Bedrooms	Bathrooms	Area	Seller_Type
	0	1	Cairo	Villa	Sale	5000000	4	3	250	Individual
	1	2	Giza	Apartment	Sale	1500000	2	1	100	Developer
	2	5	Giza	Villa	Sale	3000000	3	2	200	Individual
	3	6	Alexandria	Apartment	Sale	1200000	2	1	90	Developer

```
In [5]: # Calculate summary statistics:
         data.describe()
Out[5]:
               Property_ID
                                 Price Bedrooms Bathrooms
                                                                Area
                                                  11.000000
                                                            11.000000
         count
                 11.000000 1.100000e+01
                                       11.000000
                  9.818182 3.100000e+06
                                        3.363636
                                                  2.363636 196.363636
         mean
                  6.177672 1.560769e+06
                                        1.120065
                                                  1.120065
                                                            84.649008
           std
                  1.000000 1.200000e+06
                                        2.000000
                                                  1.000000
                                                            90.000000
          min
                  5.500000 1.700000e+06
                                                  1.500000 120.000000
          25%
                                        2.500000
                 10.000000 3.000000e+06
          50%
                                        3.000000
                                                  2.000000 200.000000
          75%
                 14.500000 4.350000e+06
                                        4.000000
                                                  3.000000 265.000000
                 19.000000 5.500000e+06
                                        5.000000
                                                  4.000000 320.000000
          max
In [6]: # Checking for null values:
         pd.isna(data).sum()
Out[6]: Property_ID
        Location
        Property_Type
                           0
        Listing Type
                           0
        Price
        Bedrooms
                          0
                           0
        Bathrooms
        Area
                           0
        Seller Type
        dtype: int64
        There is no null values
         #The shape of the data:
In [7]:
         data.shape
         (11, 9)
Out[7]:
         #Check for the duplication of the rows:
In [8]:
         data.duplicated().sum()
Out[8]:
        No duplication of the rows
In [9]:
        # checking the info of the data:
         data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 11 entries, 0 to 10
        Data columns (total 9 columns):
         # Column Non-Null Count Dtype
         ---
         0 Property_ID 11 non-null int64
1 Location 11 non-null object
         2 Property_Type 11 non-null
                                             object
         3 Listing_Type 11 non-null object
```

7

Cairo

Villa

Sale 4500000

5

300

Individual

4

4

Price

11 non-null

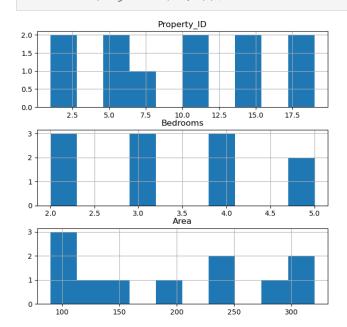
int64

5 Bedrooms 11 non-null int64 6 Bathrooms 11 non-null int64 7 Area 11 non-null int64 8 Seller_Type 11 non-null object

dtypes: int64(5), object(4)
memory usage: 920.0+ bytes

Data Exploration and Analysis and Visualization:

In [10]: # Identify trends and patterns:
 data.hist(figsize=(17,7));





In [11]: #Average Bedrooms and Bath rooms by Location and Property type
 # Calculate the average bedrooms and bathrooms by location and property type
 average_bed_bath_by_location_property = data.groupby(['Location', 'Property_Type'])[['Be
 average_bed_bath_by_location_property

Out[11]: Bedrooms Bathrooms

Location	Property_Type		
Alexandria	Apartment	2.000000	1.000000
	Villa	5.000000	4.000000
Cairo	Apartment	3.000000	2.000000
	Villa	4.333333	3.333333
Giza	Apartment	2.500000	1.500000
	Villa	3.500000	2.500000

The questions I need to answer from the data are:

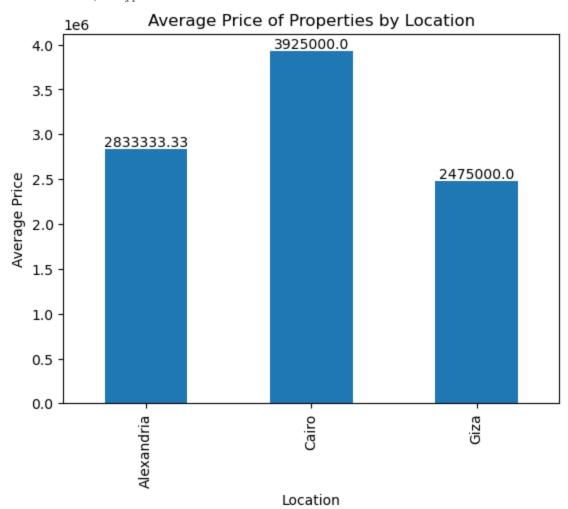
- 1- What is the average price of properties in each location?
- 2- What is the average price of properties and Property Type in each location?
- 3- Which property type (Villa or Apartment) is more common?
- 4- What is the distribution of properties based on the seller type?

```
In [25]: # Average price by location
    average_price_by_location = data.groupby('Location')['Price'].mean()
    print(average_price_by_location)

# Visualize the average price by location
    ax = average_price_by_location.plot(kind='bar')
    plt.xlabel('Location')
    plt.ylabel('Average Price')
    plt.title('Average Price of Properties by Location')

# Add data labels above each bar
    for p in ax.patches:
        ax.annotate(str(round(p.get_height(), 2)), (p.get_x() + p.get_width() / 2., p.get_he
    plt.show()
```

Location
Alexandria 2.833333e+06
Cairo 3.925000e+06
Giza 2.475000e+06
Name: Price, dtype: float64



Cairo has the highest average price among all locations.

```
In [29]: # Average price and Property type by location
  average_price_by_location_property_type = data.groupby(['Location', 'Property_Type'])['P
      print(average_price_by_location_property_type)

# Reshape the data to have Property_Type as columns and Location as index
      average_price_by_location_property_type = average_price_by_location_property_type.unstac

# Plotting the grouped bar chart
      ax = average_price_by_location_property_type.plot(kind='bar', stacked=False)
```

```
plt.xlabel('Location')
plt.ylabel('Average Price')
plt.title('Average Price by Location and Property Type')
plt.legend(title='Property Type')

for p in ax.patches:
    ax.annotate(str(p.get_height()), (p.get_x() + p.get_width() / 2., p.get_height()), h

plt.show()

Location    Property_Type

The reduction    Property_Type

The reduction    Property_Type
```

Name: Price, dtype: float64



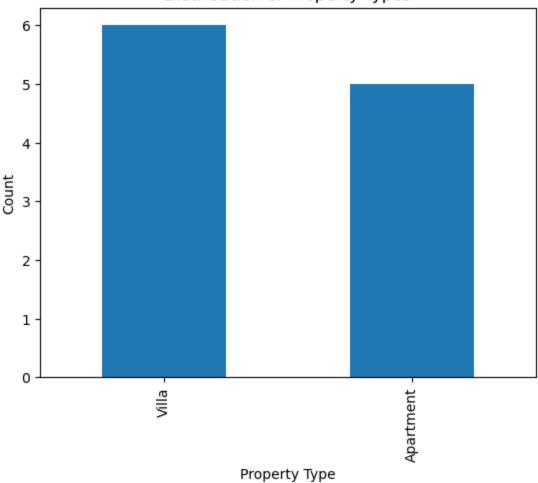
The average price of villas is the highest among all property types and locations, and the average price of apartments in Cairo is the highest among all locations.

```
In [30]: # Count the occurrences of each property type
property_type_counts = data['Property_Type'].value_counts()
print(property_type_counts)

# Plotting the bar chart
property_type_counts.plot(kind='bar')
plt.xlabel('Property Type')
plt.ylabel('Count')
plt.title('Distribution of Property Types')
plt.show()
```

Villa 6
Apartment 5
Name: Property Type, dtype: int64

Distribution of Property Types



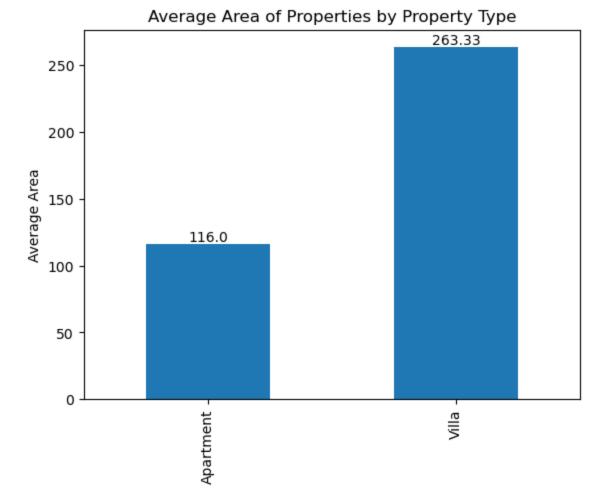
The count of Villas in the data is greater than the count of Apartments in the data.

```
In [31]: # Average area of properties by property type
    average_area_by_property_type = data.groupby('Property_Type')['Area'].mean()
    print(average_area_by_property_type)

# Plotting the bar chart
    ax = average_area_by_property_type.plot(kind='bar')
    plt.xlabel('Property Type')
    plt.ylabel('Average Area')
    plt.title('Average Area of Properties by Property Type')

for p in ax.patches:
    ax.annotate(str(round(p.get_height(), 2)), (p.get_x() + p.get_width() / 2., p.get_he
    plt.show()
```

Property_Type
Apartment 116.000000
Villa 263.333333
Name: Area, dtype: float64



The average area of apartments is 116 square meters, and the average area of villas is 263.33 square meters.

Property Type

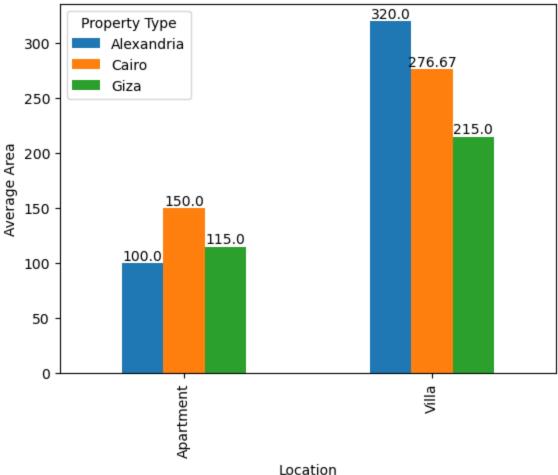
```
In [32]:
         #the average area of properties by property type and location:
         average_area_by_property_type_location = data.groupby(['Property Type', 'Location'])['Ar
        print(average area by property type location)
         # Reshape the data to have Property Type as columns and Location as index
        average area by property type location = average area by property type location.unstack(
         # Plotting the grouped bar chart
        ax = average area by property type location.plot(kind='bar')
        plt.xlabel('Location')
        plt.ylabel('Average Area')
        plt.title('Average Area of Properties by Property Type and Location')
        plt.legend(title='Property Type')
         # Add data labels above each bar
         for p in ax.patches:
            ax.annotate(str(round(p.get height(), 2)), (p.get x() + p.get width() / 2., p.get he
        plt.show()
        Property Type Location
                                     100.000000
        Apartment
                       Alexandria
                                     150.000000
                       Cairo
                       Giza
                                     115.000000
        Villa
                       Alexandria
                                     320.000000
                       Cairo
                                     276.666667
```

215.000000

Giza

Name: Area, dtype: float64

Average Area of Properties by Property Type and Location



The highest average area is observed for villas in Alexandria, and the highest average area is observed for apartments in Cairo.

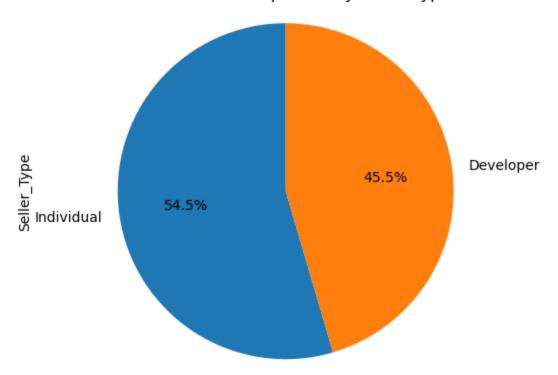
```
In [33]: # Count the properties by seller type
    seller_type_counts = data['Seller_Type'].value_counts()
    print(seller_type_counts)

# Visualize the distribution of properties by seller type
    seller_type_counts.plot(kind='pie', autopct='%1.1f%%', startangle=90)
    plt.title('Distribution of Properties by Seller Type')
    plt.axis('equal')
    plt.show()

Individual 6
```

Developer 5
Name: Seller_Type, dtype: int64

Distribution of Properties by Seller Type



The distribution of seller types shows that there is a higher proportion of individual sellers compared to developers.

About Rent Data:

In [18]: # Read the dataset
Data = pd.read_excel(r"data test Rent.xlsx")
Data

Out[18]:		Property_ID	Location	Property_Type	Listing_Type	Price	Bedrooms	Bathrooms	Area	Seller_Type
	0	3	Alexandria	Apartment	Rent	8000	1	1	80	Individual
	1	4	Cairo	Apartment	Rent	10000	2	2	120	Developer
	2	8	Giza	Apartment	Rent	7000	1	1	70	Individual
	3	9	Alexandria	Apartment	Rent	9000	2	1	100	Developer
	4	12	Alexandria	Apartment	Rent	7500	1	1	85	Individual
	5	13	Cairo	Apartment	Rent	8500	2	2	110	Developer
	6	16	Cairo	Apartment	Rent	9500	2	1	95	Individual
	7	17	Giza	Apartment	Rent	7500	1	1	75	Developer
	8	20	Giza	Apartment	Rent	6500	1	1	60	Individual

In [19]: # Calculate summary statistics
 Data.describe()

Property_ID Out[19]: **Price Bedrooms Bathrooms** Area 9.000000 9.000000 9.000000 9.000000 9.000000 count mean 11.333333 8166.666667 1.444444 1.222222 88.333333

std	5.830952	1172.603940	0.527046	0.440959	19.525624
min	3.000000	6500.000000	1.000000	1.000000	60.000000
25%	8.000000	7500.000000	1.000000	1.000000	75.000000
50%	12.000000	8000.000000	1.000000	1.000000	85.000000
75%	16.000000	9000.000000	2.000000	1.000000	100.000000
max	20.000000	10000.000000	2.000000	2.000000	120.000000

In [20]: #Average Bedrooms and Bath rooms by Location and Property type
Calculate the average bedrooms and bathrooms by location and property type
average_bed_bath_by_location_property = Data.groupby(['Location'])[['Bedrooms', 'Bathrooms']])
average_bed_bath_by_location_property

Out[20]: Bedrooms Bathrooms

Location

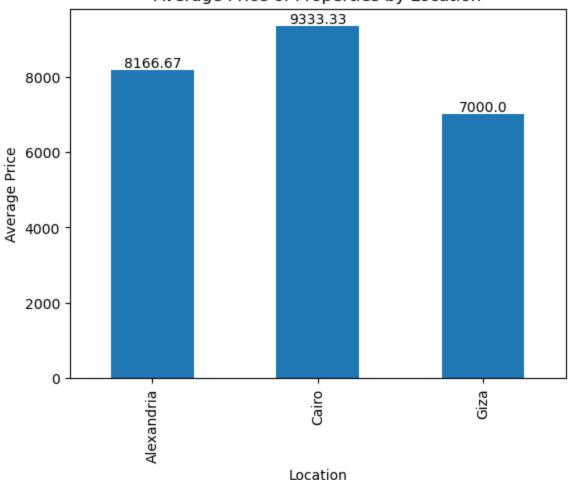
Alexandria	1.333333	1.000000
Cairo	2.000000	1.666667
Giza	1.000000	1.000000

```
In [21]: # Calculate the average price by location
    average_price_by_location = Data.groupby('Location')['Price'].mean()

# Plotting the bar chart
    ax = average_price_by_location.plot(kind='bar')
    plt.xlabel('Location')
    plt.ylabel('Average Price')
    plt.title('Average Price of Properties by Location')

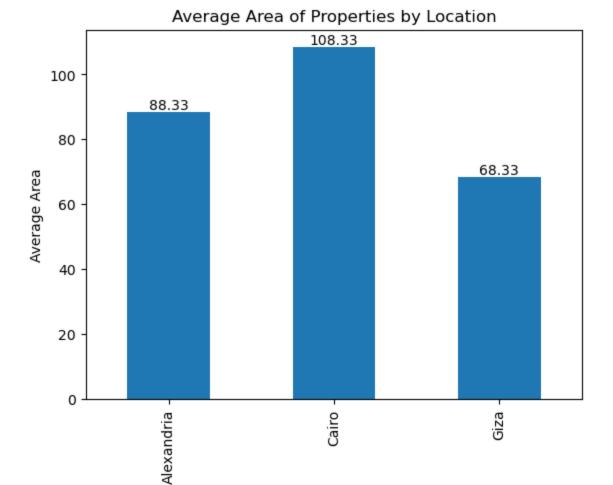
for p in ax.patches:
    ax.annotate(str(round(p.get_height(), 2)), (p.get_x() + p.get_width() / 2., p.get_he
    plt.show()
```

Average Price of Properties by Location



The average Rent Price in Cairo is the highest

Alexandria 88.333333 Cairo 108.333333 Giza 68.333333 Name: Area, dtype: float64



the average area of apartments for rent in Cairo is the highest compared to other locations.

Name: Seller Type, dtype: int64

```
In [35]: # Count the properties by seller type
property_count_by_seller_type = Data['Seller_Type'].value_counts()

print(property_count_by_seller_type)

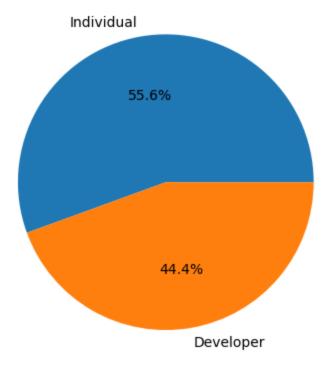
# Plotting the pie chart
plt.pie(property_count_by_seller_type, labels=property_count_by_seller_type.index, autop plt.title('Count of Properties by Seller Type')

plt.show()

Individual 5
Developer 4
```

Location

Count of Properties by Seller Type



The distribution of seller types shows that there is a higher proportion of individual sellers compared to developers.





Insights For Sales Data:

1- Cairo has the highest prices among all locations and the average price of apartments in Cairo is the highest among all locations and The selling price in Cairo from individuals is the highest, regardless of the area.

In my opinion, the reasons are as follows:

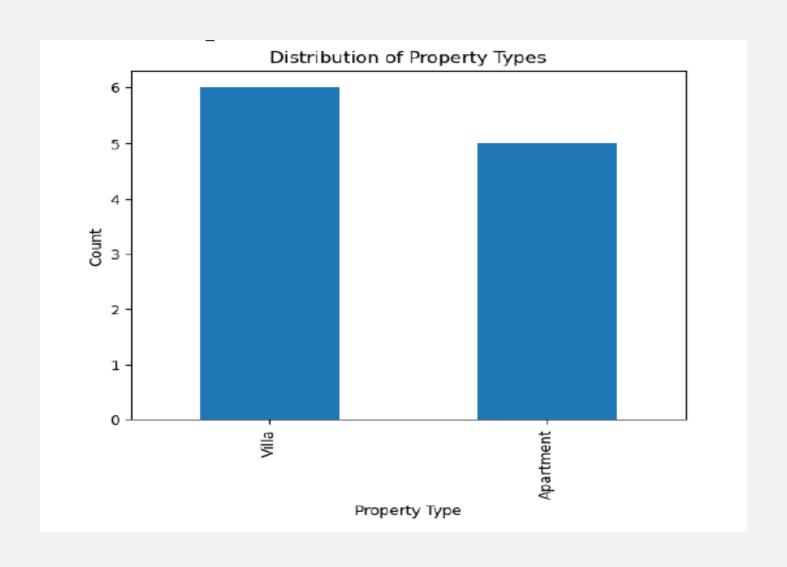
- **1-Economic Importance:** Cairo is the capital city of Egypt and serves as the country's political, administrative, and economic center. Being the heart of economic activities, Cairo attracts significant investments, businesses, and job opportunities. This economic prominence can drive up demand for properties in Cairo, leading to higher sale prices.
- **2-Population Density:** Cairo has a much larger population compared to Alexandria and Giza. With a population of over 20 million people, Cairo is densely populated, which creates a high demand for housing and limited availability of land. The scarcity of land and high population density can result in higher property prices.
- **3-Infrastructure and Amenities:** Cairo generally offers better infrastructure and a wider range of amenities compared to other cities. It has a well-developed transportation system, including a metro network, which makes commuting easier. Cairo also boasts a variety of educational institutions, healthcare facilities, shopping centers, and entertainment options. The availability of these amenities and infrastructure can contribute to higher property prices.
- **4-Employment Opportunities:** As the capital city, Cairo offers numerous job opportunities across various industries, including government, finance, media, and tourism. Many individuals migrate to Cairo in search of better employment prospects, leading to increased demand for housing. The higher demand for properties can drive up sale prices.
- **5-Real Estate Market Dynamics:** The dynamics of the real estate market, including supply and demand, investor preferences, and market speculation, can influence property prices. Cairo's real estate market may have experienced significant growth and investor interest, leading to higher sale prices. Additionally, factors such as urbanization, gentrification, and development projects can also impact property values.

2- The average price of villas is the highest among all property types and locations.

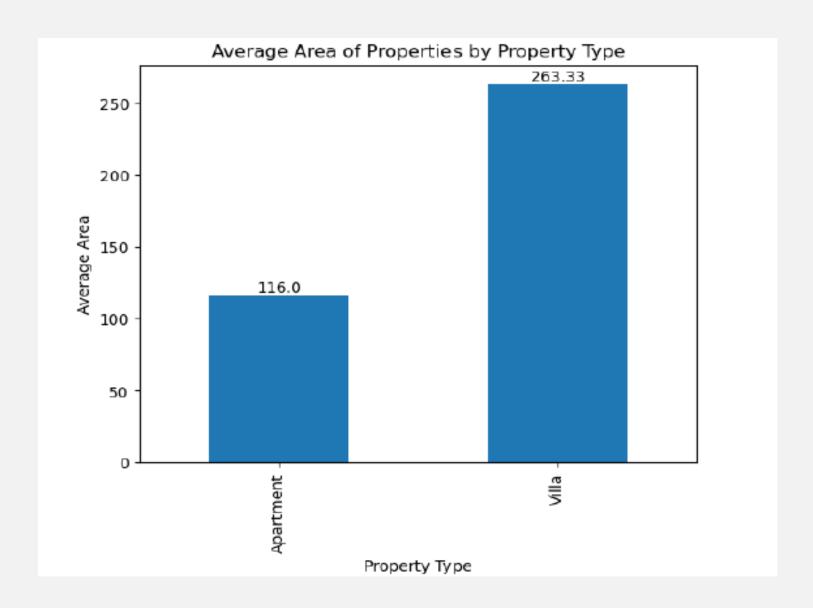
In my opinion, the reasons are as follows:

- **1- Demand and Prestige:** Villas are often associated with luxury, spaciousness, and exclusivity. They tend to offer more privacy, larger living spaces, and amenities such as gardens, pools, and private parking. The higher average price of villas may be due to the high demand for such prestigious and upscale properties.
- **2- Scarcity of Land:** Villas typically require larger plots of land compared to other property types. In densely populated areas like cities, finding available land for villas can be challenging. The limited supply of land suitable for villas can drive up their average prices.
- **3- Location and Desirability:** The location of villas can significantly influence their prices. Villas situated in prime or sought-after locations, such as exclusive neighborhoods, waterfronts, or areas with scenic views, tend to command higher prices. Factors like proximity to amenities, good infrastructure, and safety can also contribute to the desirability and higher prices of villas.
- **4- Quality and Customization:** Villas often offer high-quality construction materials, finishes, and customization options. Buyers may be willing to pay a premium for the superior craftsmanship, unique architectural designs, and personalized features that villas offer.

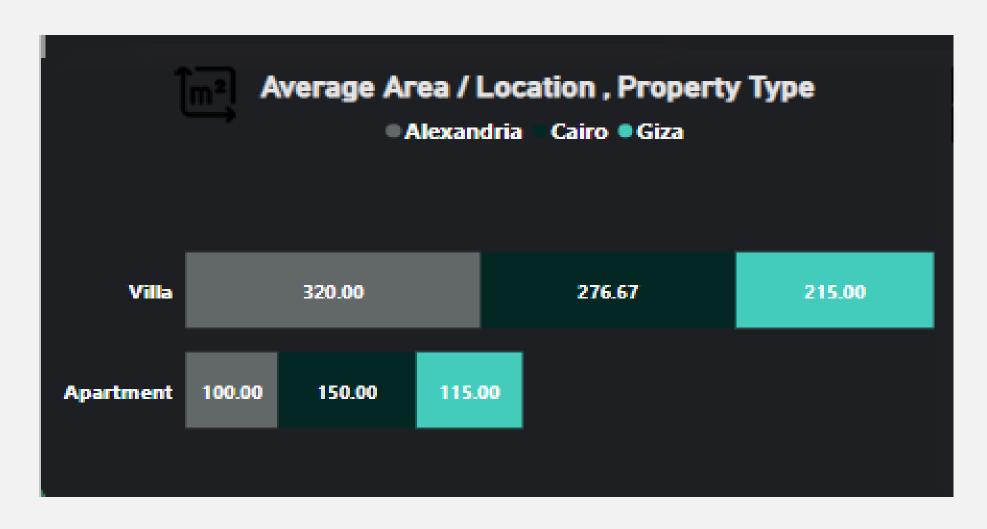
3- The count of villas in the data is greater than the count of apartments.



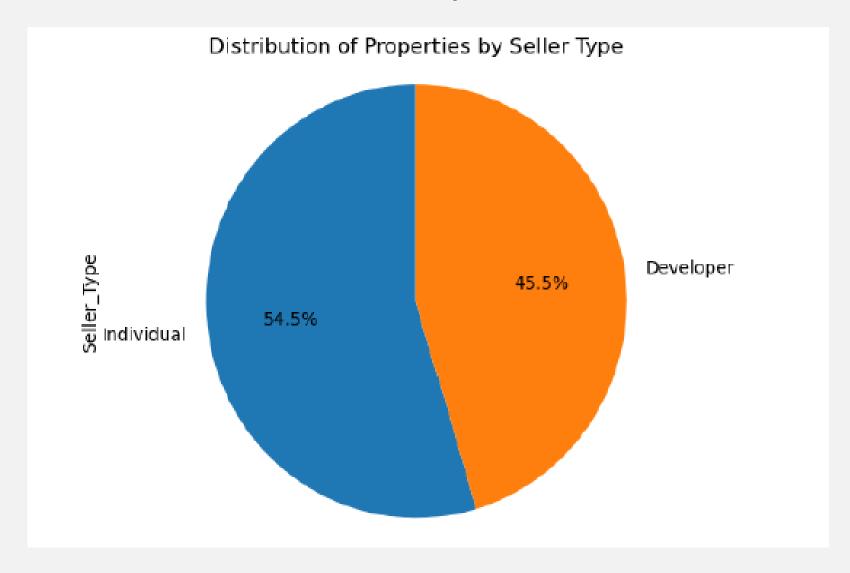
4- The average area of apartments is 116 square meters, and the average area of villas is 263.33 square meters.



5- The highest average area is observed for villas in Alexandria, and the highest average area is observed for apartments in Cairo.



6- The distribution of seller types shows a higher proportion of individual sellers compared to developers.



7- There is a positive correlation between the selling price in Alexandria and Giza and the area, regardless of the seller type.

In my opinion, the reasons are as follows:

- 1- Supply and Demand: Larger properties, typically associated with larger areas, are often in higher demand due to the benefits they offer, such as more living space, potential for expansion, and increased privacy. The limited supply of larger properties compared to smaller ones can drive up their prices.
- **2- Increased Utility:** Larger properties provide more usable space, both indoors and outdoors. This can be particularly appealing to buyers who have specific needs, such as families requiring more bedrooms, space for home offices, or outdoor areas for recreational activities.
- **3- Land Value:** The value of land typically increases with its size, especially in desirable areas. As the area of a property increases, so does the land area associated with it. If the property is located in a sought-after neighborhood or has desirable features, the land value can significantly impact the sale price.
- **4- Construction Costs:** Generally, constructing a larger property requires more resources, materials, and labor, resulting in higher construction costs. The sale price of a property often reflects these costs, along with a markup for profit. Therefore, larger properties tend to have higher sale prices to account for the higher construction expenses.
- **5- Market Preferences:** Buyer preferences and trends can influence property prices. In some markets, there may be a preference for larger properties, which can drive up their prices. This could be due to cultural factors, lifestyle preferences, or changing market dynamics.

8- The selling prices in Cairo are generally higher from individuals than from developers.

In my opinion, the reasons are as follows:

Location: Individual sellers or landlords might own properties in prime locations within Cairo, such as in the city center or in popular neighborhoods. These locations often command higher prices due to their proximity to amenities, transportation, and commercial centers.

Flexibility: Individual sellers or landlords may have more flexibility in negotiating prices and rental terms compared to developers who often have fixed pricing structures. This flexibility allows individuals to set higher prices based on factors such as demand, property condition, or unique features.

Supply and demand: In some cases, the supply of properties from developers might outstrip the demand, leading to more competitive pricing and lower rental rates. On the other hand, individual sellers may have a limited number of properties available, creating a situation where demand exceeds supply and drives up prices.

Property condition and customization: Individual sellers may have invested in renovations, upgrades, or customization of their properties, which can justify higher selling or rental prices. Buyers may be willing to pay more for properties that offer desirable features, modern amenities, or unique design elements.

Middleman fees: Individuals often involve real estate agencies or brokers in the sale or rental process, which can add additional fees and commissions to the overall cost. Dealing directly with developers can sometimes eliminate these middleman fees, resulting in lower overall costs for buyers.

Insights For Rent Data:

- 1- Cairo has the highest prices among all locations.
- 2- The average area of apartments in Cairo is the highest among all locations.
- 3- The distribution of seller types shows a higher proportion of individual sellers compared to developers.
- 4- The rental price in Cairo from individuals is the highest, regardless of the area.
- 5- There is a direct correlation between the selling price in Alexandria and Giza and the area, regardless of the seller type.
- 6- The rental prices in Cairo are generally higher from individuals than from developers.

