

**SOUTHERN ALBERTA INSTITUTE OF TECHNOLOGY**

**SCHOOL FOR ADVANCED DIGITAL TECHNOLOGY (SADT)**

**FINAL VISUALIZATION PROJECT REPORT**  
**AN ANALYSIS OF HOUSING PRICE & REAL ESTATE IN 2023**

**By**

**GROUP 9**

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## **Goal:**

The goal of the project is to study the Canadian housing market in 2023 to offer information on potential investment opportunities in different regions and types of properties. Using data analysis methods and visual representations, the project aims to grasp trends in housing prices, the impact of demographics.

## **Datasets Used & Licensing References:**

### **1. Housing Price & Real Estate - 2023:**

- Data Source: Kaggle
- Licensing: Public Domain
- URL: <https://www.kaggle.com/datasets/reenapinto/housing-price-and-real-estate-2023>
- Description: This dataset provides information about housing prices and real estate trends in Canada for the year 2023. It includes data such as property prices, descriptions, locations, and other relevant attributes.

### **2. Community Points:**

- Data Source: Calgary Open Data
- Licensing: Open Government License - City of Calgary
- URL: <https://data.calgary.ca/Base-Maps/Community-Points/j9ps-fyst/data>
- Description: This dataset contains community points in Calgary, where each point identifies the centroid of a specific community. It includes geographic coordinates and other community-related attributes.

## **Citations:**

- Kaggle Housing Price & Real Estate - 2023: Reena (Owner), Kaggle
- Calgary Open Data - Community Points: The City of Calgary, Calgary Open Data

By utilizing these datasets and adhering to the respective licensing terms, the project aims to provide comprehensive insights into the Canadian housing market in 2023, facilitating informed decision-making for investors.

## **Introduction:**

- In this study, we analyze the Canadian housing market in 2023 using data from reputable sources like Kaggle and Calgary Open Data.
- Our objective is to provide actionable insights for stakeholders.
- We follow the CRISP-DM methodology, beginning with a thorough understanding of the business context and data.
- We formulate relevant business questions and create visualizations to address them effectively.
- Our findings are presented through a Power BI report, offering a comprehensive view of housing trends and investment opportunities.

## **1. Business Understanding**

### **Objective:**

- The objective is to analyze trends in the housing market, particularly focusing on information on potential investment opportunities in different regions and types of properties.

### **Background:**

- Housing market experts believe there's a shift towards a seller's market due to increased demand and limited supply.
- Factors contributing to increased demand include millennials entering the market as first-time buyers.
- Both residential and commercial sectors are experiencing heightened demand.
- Some experts speculate that this demand surge might be contributing to a housing crisis.
- Canada is implementing measures like capping immigration to address potential housing issues.

### **Stakeholders:**

- Real estate agencies, Homeowners, Prospective homebuyers, Property investors

### **Success Criteria:**

- Ability to provide insights into factors driving the market (e.g., demographics, economic trends, housing supply).

### **Constraints:**

- Availability and quality of data: Ensuring access to reliable data sources for analysis.
- Regulatory factors: Compliance with real estate regulations and laws.

### **Assumptions:**

- The analysis is based on 2023 housing prices listed on various websites, trends and may change depending on future economic and social developments.

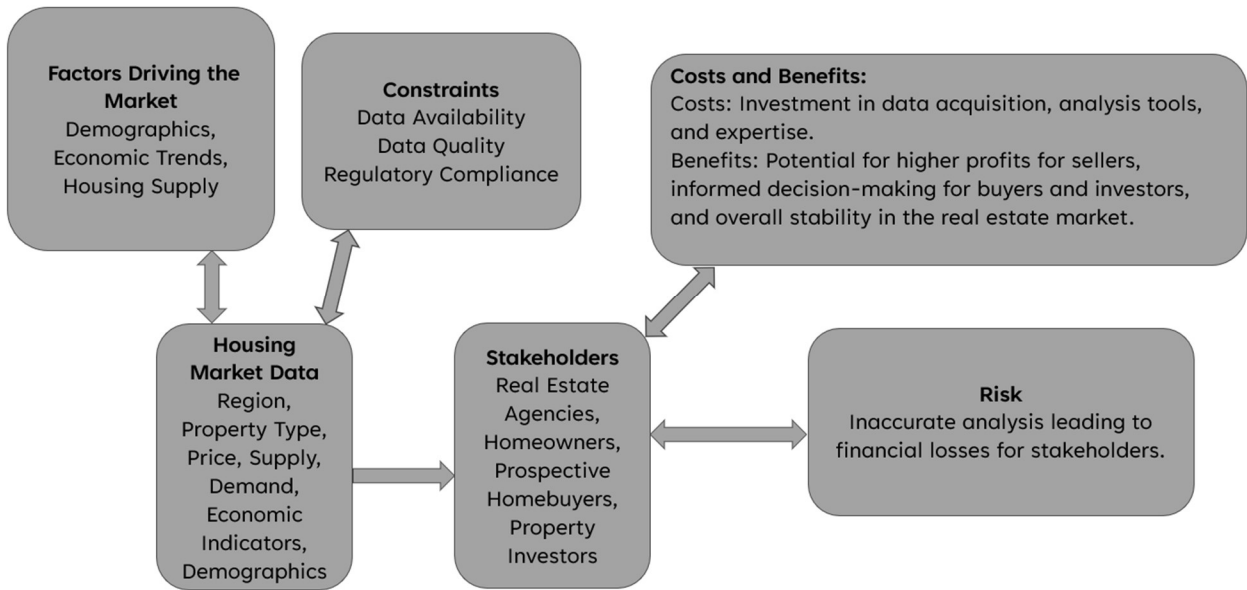
Risks and Contingencies:

- Risk: Inaccurate analysis leading to financial losses for stakeholders.
- Contingency: Regular monitoring and updating of models based on new data and market changes.

Costs and Benefits:

- Costs: Investment in data acquisition, analysis tools, and expertise.
- Benefits: Potential for higher profits for sellers, informed decision-making for buyers and investors, and overall stability in the real estate market.

Conceptual Model (Data Map):



Data Dictionary:

Topic	Definition
Region	Refers to a geographical area within Canada, such as provinces, cities, or neighborhoods, where housing market data is collected and analyzed.
Property Type	Describes the category or classification of a property, such as single-family homes, condominiums, or commercial properties, which influences its market value and demand.
Demand	Refers to the level of interest or desire for properties in a specific region or property type, influenced by factors such as population growth, economic conditions, and buyer preferences.
Supply	Indicates the availability of properties for sale or rent within a given region or property type, impacting market dynamics, pricing, and competition among buyers and sellers.
Regulatory Compliance	Involves adherence to real estate regulations, laws, and ethical standards governing data collection, analysis, and reporting to ensure legality, transparency, and ethical conduct in the housing market.
Demographics	Encompasses the statistical characteristics of a population within a region, including age, income, household composition, and cultural diversity, which influence housing demand and market trends.

## 2. Data Understanding

- Housing Price & Real Estate 2023: Contains data on house listings, including price, geographic location, property type, and additional features influencing housing value.

<pre># Import the necessary library and load the datasets import pandas as pd  # Load Homes for Sale and Real Estate data set Housing = pd.read_excel("Homes_for_Sale_and_Real_Estate.xlsx") Housing.head(5)</pre>							
✓ 0.3s							
	Address	Price	Description	Place	Beds	Bath	Sq.Ft
0	3704 42 St SW	979999	CA AB T3E 3N1	Glenbrook	4	3.5	1813
1	30 Mahogany Mews SE #415	439900	CA AB T3M 3H4	Mahogany	2	2.0	1029
2	273 Auburn Shores Way SE	950000	CA AB T3M 2E9	Auburn Bay	4	2.5	2545
3	235 15 Ave SW #404	280000	CA AB T2R 0P6	Beltline	2	2.0	898
4	24 Hemlock Crescent SW #2308	649000	CA AB T3C 2Z1	Spruce Cliff	2	2.0	1482

```
✓ Housing.info() ...

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3360 entries, 0 to 3359
Data columns (total 8 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Address     3360 non-null   object
1   Price       3360 non-null   int64
2   Description 3360 non-null   object
3   Place       3344 non-null   object
4   Beds        3360 non-null   int64
5   Bath        3360 non-null   float64
6   Sq.Ft       3360 non-null   int64
7   Website     3359 non-null   object
dtypes: float64(1), int64(3), object(4)
memory usage: 210.1+ KB
```

## 3 & 4. Data Preparation: Cleaning and Modeling:

Clean and preprocess the datasets to handle missing values, outliers, and inconsistencies.

1. Checked blank rows and printed them. There were empty values in the Place and Website columns. As our analysis relied on the Place data, it was necessary to retain these rows.

<pre># Check for blank rows blank_rows_housing = Housing[Housing.isnull().any(axis=1)]  # Display the blank rows, if any if not blank_rows_housing.empty:     print("Blank rows found in Housing DataFrame:")     print(blank_rows_housing) else:     print("No blank rows found in Housing DataFrame.")</pre>							
Blank rows found in Housing DataFrame:							
	Address	Price	Description	Place	Beds	Bath	\
122	1066 Creekside Blvd SW	580900	CA AB T2X5K6	NaN	1	1.5	
160	341 Walcrest View SE	820000	CA AB T2X 4V9	Walden	5	3.5	
641	62 Royston Terrace NW	849900	CA AB T3L 0J2	NaN	3	2.5	
1154	3250 84 St SE	2400000	CA AB T2B 3C1	NaN	6	2.0	
1174	8535 19 Ave SE #424	455000	CA AB T2A 7M8	NaN	2	1.5	
1245	99 Taralake Way NE	672000	CA AB T3J0A7	NaN	5	3.5	
1324	148 Savanna Dr NE	850000	CA AB T3J2H5	NaN	4	3.0	
1519	9110 34 Ave	2299000	CA AB T1X 0L5	NaN	6	4.5	
1630	71 Lynx Meadows Dr NW	2000000	CA AB T3L 3L9	NaN	6	4.5	
1846	8535 19 Ave SE #421	505000	CA AB T2A 7M8	NaN	3	2.5	
1867	4520 84 Ave NE	1000000	CA AB T3J 4C4	NaN	5	4.0	
2019	99 Royston Rise NW	794900	CA AB T3L 0J2	NaN	3	2.5	
2361	1161 Creekside Blvd SW	667500	CA AB T2X5K5	NaN	3	2.5	
2629	8535 19 Ave SE #417	488900	CA AB T2A 7M8	NaN	4	3.5	
3033	30 Forzani Way NW	2388000	CA AB T3Z 1L5	NaN	5	3.5	
3311	902 Bluerock Way SW	702500	CA AB T2Y 0S5	NaN	3	2.5	
3356	4111 162 Ave SW	8000000	CA AB T2Y 0N7	NaN	5	4.5	
	Sq.Ft		Website				
122	964		Maxwell Canyon Creek				
160	2235		NaN				
641	2467		Cir Realty				
1154	2147		RE/MAX Key				
...							
2629	1365		Exp Realty				
3033	2554		Cir Realty				
3311	2096		Bode				
3356	9031		Diamond Realty & Associates Ltd.				

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- To preserve the rows, I decided to replace them with text extracted from the address column.

```
###

import re

# Function to extract text from address
def extract_text(address):
    return re.sub(r'\d+', '', address).strip()

# Apply the function to create a new column with only text from 'Address'
Housing['TextAddress'] = Housing['Address'].apply(extract_text)

# Store rows where 'Place' is blank before replacement
blank_rows_before = Housing[Housing['Place'].isnull()]

# Replace blank rows in 'Place' column with text from 'TextAddress' column
Housing['Place'] = Housing['Place'].fillna(Housing['TextAddress'])

# Store rows where 'Place' was replaced
replaced_rows = Housing.loc[blank_rows_before.index]

# Drop the 'TextAddress' column if you don't need it anymore
Housing.drop(columns=['TextAddress'], inplace=True)

# Print replaced rows
print("Replaced Rows:")
print(replaced_rows)
```

Replaced Rows:

	Address	Price	Description	Place \
122	1066 Creekside Blvd SW	580900	CA AB T2X5K6	Creekside Blvd SW
641	62 Royston Terrace NW	849900	CA AB T3L 0J2	Royston Terrace NW
1154	3250 84 St SE	2400000	CA AB T2B 3C1	St SE
1174	8535 19 Ave SE #424	455000	CA AB T2A 7W8	Ave SE #
1245	99 Taralake Way NE	672000	CA AB T3J0A7	Taralake Way NE
1324	148 Savanna Dr NE	850000	CA AB T3J2H5	Savanna Dr NE
1519	9110 34 Ave	2299000	CA AB T1X 0L5	Ave
1630	71 Lynx Meadows Dr NW	2000000	CA AB T3L 3L9	Lynx Meadows Dr NW
1846	8535 19 Ave SE #421	505000	CA AB T2A 7W8	Ave SE #
1867	4520 84 Ave NE	1000000	CA AB T3J 4C4	Ave NE
2019	99 Royston Rise NW	794900	CA AB T3L 0J2	Royston Rise NW
2361	1161 Creekside Blvd SW	667500	CA AB T2X5K5	Creekside Blvd SW
2629	8535 19 Ave SE #417	488900	CA AB T2A 7W8	Ave SE #
3033	30 Forzani Way NW	2388000	CA AB T3Z 1L5	Forzani Way NW
3311	902 Bluerock Way SW	702500	CA AB T2Y 0S5	Bluerock Way SW
3356	4111 162 Ave SW	8000000	CA AB T2Y 0N7	Ave SW

### 3. Import Data

Then, we imported data into power bi using import data from csv option.

Comma-Separated Values

☒ Basic ☐ Advanced

File path

Open file as

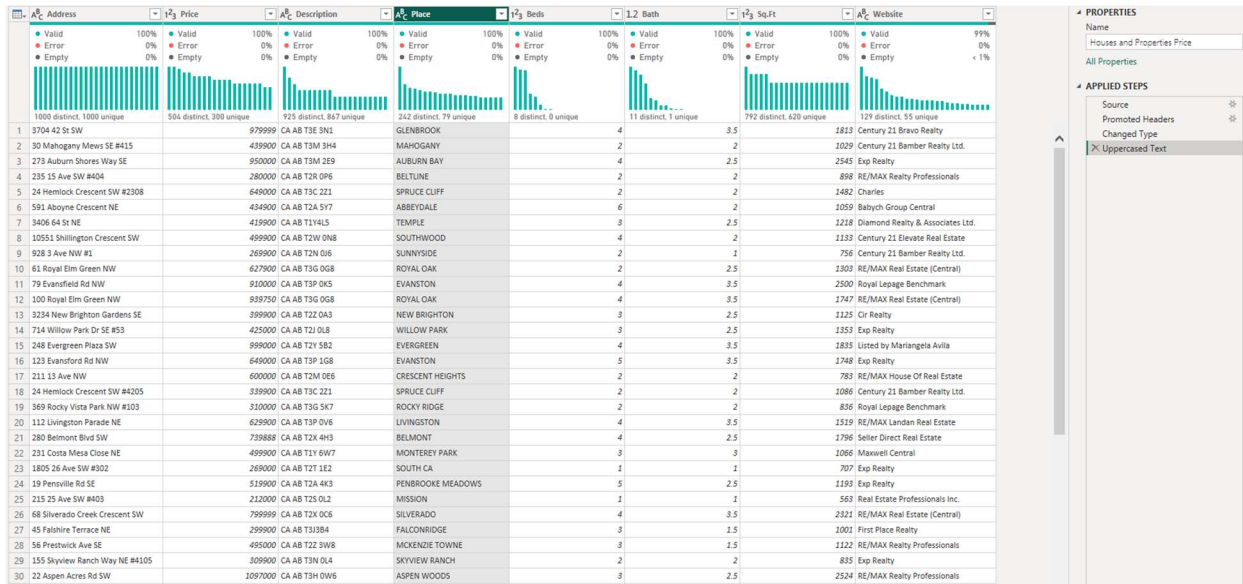
File origin

Line breaks

Delimiter

#### 4. Cleaning the data and transforming the data

- Checked to see if we have any unnecessary columns since all columns were relevant in our case, we decided to keep all the columns.
- Correcting Errors: There were no errors to fix when we loaded the data.
- We made sure the data types were appropriate for the data set and columns were named appropriately.



#### 5. We created a new column with price per square foot.

1 Price per Sq.ft. = 'Houses and Properties Price'[Price] / 'Houses and Properties Price'[Sq.Ft]							
Address	Price	Description	Place	Beds	Bath	Sq.Ft	Website
30 Mahogany Mews SE #415	\$439,900	CA AB T3M 3H4	MAHOGANY	2	2	1029	Century 21 Bamber Realty Ltd.
235 15 Ave SW #404	\$280,000	CA AB T2R 0P6	BELTLINE	2	2	898	RE/MAX Realty Professionals
24 Hemlock Crescent SW #2308	\$649,000	CA AB T3C 2Z1	SPRUCE CLIFF	2	2	1482	Charles
211 13 Ave NW	\$600,000	CA AB T2M 0E6	CRESCENT HEIGHTS	2	2	783	RE/MAX House Of Real Estate
24 Hemlock Crescent SW #4205	\$339,900	CA AB T3C 2Z1	SPRUCE CLIFF	2	2	1086	Century 21 Bamber Realty Ltd.
369 Rocky Vista Park NW #103	\$310,000	CA AB T3G 5K7	ROCKY RIDGE	2	2	836	Royal Lepage Benchmark
155 Skyview Ranch Way NE #4105	\$309,900	CA AB T3N 0L4	SKYVIEW RANCH	2	2	835	Exp Realty
1107 Gladstone Rd NW #503	\$828,345	CA AB T2M3T1	HILLHURST	2	2	1094	Bode Platform Inc.
10 Discovery Ridge Close SW #412	\$539,000	CA AB T3H 5X3	DISCOVERY RIDGE	2	2	1379	Century 21 Powerrealty.Ca
99 Copperstone Park SE #3415	\$299,900	CA AB T2Z 5C9	COPPERFIELD	2	2	803	Royal Lepage Mission Real Estate
77 Spruce Place SW #2607	\$539,800	CA AB T3C3X6	SPRUCE CLIFF	2	2	1148	RE/MAX Complete Realty
1010 Arbour Lake Rd NW #1306	\$363,900	CA AB T3G 4Y8	ARBOUR LAKE	2	2	985	RE/MAX Real Estate (Mountain View)
1025 5 Ave SW #1804	\$578,000	CA AB T2P 1N4	DOWNTOWN WEST END	2	2	823	Sotheby's International Realty Canada
155 Skyview Ranch Way NE #6212	\$319,000	CA AB T3N 0L1	SKYVIEW RANCH	2	2	850	The Real Estate Company
544 Blackthorn Rd NE #1302	\$335,000	CA AB T2K 5J5	THORNCIFFE	2	2	948	Bow Realty
1410 2 St SW #407	\$320,000	CA AB T2R 1R1	BELTLINE	2	2	832	Real Estate Professionals Inc.
30 Brentwood Common NW #1013	\$499,999	CA AB T2L 2L8	BRENTWOOD	2	2	753	Exp Realty
2330 Fish Creek Blvd SW #1346	\$859,900	CA AB T2Y 0L1	EVERGREEN	2	2	1367	RE/MAX Realty Innovations
10 Auburn Bay Ave SE #513	\$449,900	CA AB T3M 0P8	CHESAPEAKE	2	2	1249	Engel & Volkers CA
88 9 St NE #103	\$729,900	CA AB T2E 4E1	BRIDGELAND/RIVERSIDE	2	2	1227	Charles

We then changed the formatting of Price and Price per Sq. ft. to currency with 2 decimal places.

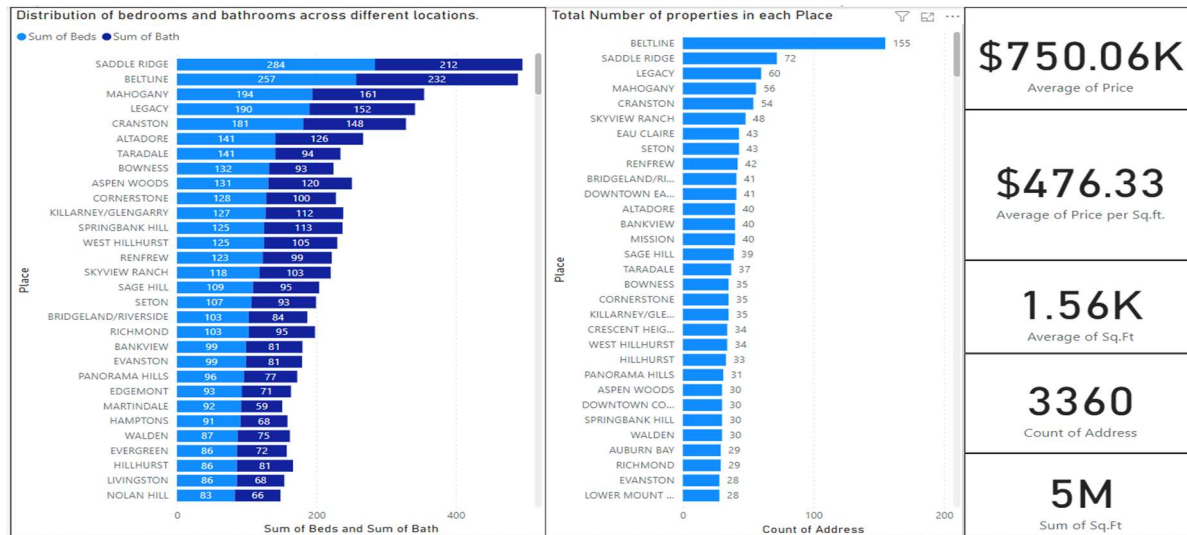
#### 5. Evaluation:

Before proceeding to deployment, we conducted a thorough review of the entire process. We did initial analysis on the data set which can be seen below,



## Initial Analysis:

From the below Dashboard we can see the Distribution of bedrooms and bathrooms across different locations, Total Number of properties in each place. Average of Price, Average of Price per Sq. ft. Average of Sq. Ft and total number of properties listed and total number of Sq. Ft which is 5M if u convert that in square kilometer it's just approx. 0.5 km<sup>2</sup> and total land area of Canada is 825.29 km<sup>2</sup> which is not even 1 percent of total land area.



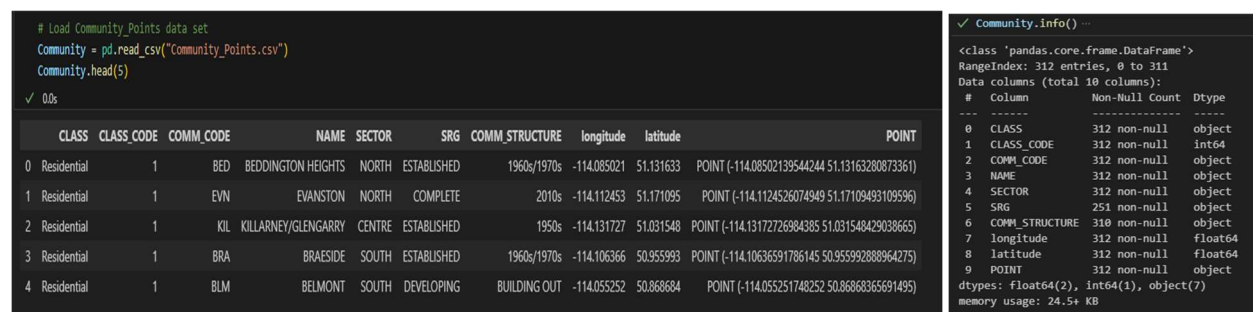
Now let's outline the next steps, since our current dataset lacks geographical data for map visualization, as well as information on sectors and community types, we need to revisit the Business Understanding phase. We will utilize the second dataset and reiterate the CRISP-DM process to address these gaps.

## 6. Business Understanding

**Objective:** The goal of the project is to study the Canadian housing market in 2023 to offer information on potential investment opportunities in different regions and types of properties. Using data analysis methods and visual representations, the project aims to grasp trends in housing prices, the impact of demographics.

## 7. Data Understanding

- Community Points: Provides geographic and demographic data for communities, including centroids for precise location mapping, community structure, and developmental stage.





## 8 & 9. Data Preparation: Cleaning and Modeling:

1. We needed latitude and longitude to work with maps, so we downloaded the Community points dataset from source 2 and checked for blank rows.

```
# Check for blank rows
blank_rows_community = Community[Community.isnull().any(axis=1)]

# Display the blank rows, if any
if not blank_rows_community.empty:
    print("Blank rows found in Housing DataFrame:")
    print(blank_rows_community)
else:
    print("No blank rows found in Housing DataFrame.")
```

Blank rows found in Housing DataFrame:

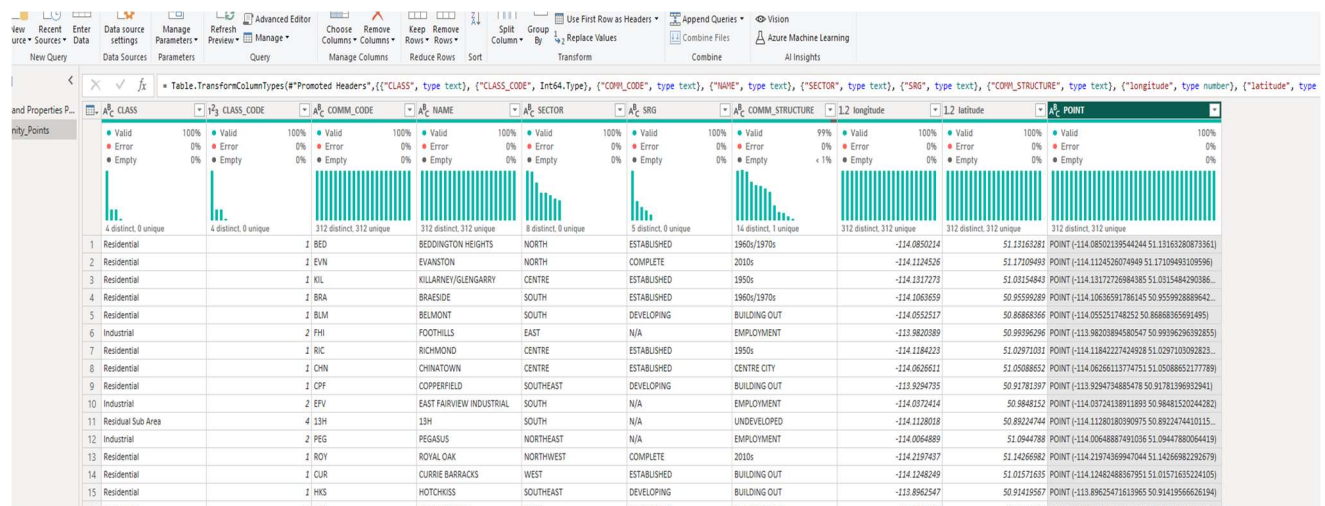
	CLASS	CLASS_CODE	COMM_CODE	NAME	\
5	Industrial	2	FHI	FOOTHILLS	
9	Industrial	2	EFV	EAST FAIRVIEW INDUSTRIAL	
10	Residual Sub Area	4	13H		13H
11	Industrial	2	PEG	PEGASUS	
16	Residual Sub Area	4	13N		13N
...	...	...	...	...	...
290	Industrial	2	MNI	MANCHESTER INDUSTRIAL	
296	Industrial	2	HIF	HIGHFIELD	
299	Residual Sub Area	4	13I		13I
300	Residual Sub Area	4	11B		11B
309	Major Park	3	FPK	FISH CREEK PARK	

	SECTOR	SRG	COMM_STRUCTURE	longitude	latitude	\
5	EAST	NaN	EMPLOYMENT	-113.982039	50.993963	
9	SOUTH	NaN	EMPLOYMENT	-114.037241	50.984815	
10	SOUTH	NaN	UNDEVELOPED	-114.112802	50.892247	
11	NORTHEAST	NaN	EMPLOYMENT	-114.006489	51.094479	
16	SOUTH	NaN	OTHER	-114.143768	50.944846	
...	...	...	...	...	...	...
290	CENTRE	NaN	EMPLOYMENT	-114.057225	51.013261	
296	CENTRE	NaN	EMPLOYMENT	-114.037201	51.018532	
299	SOUTH	NaN	UNDEVELOPED	-114.099428	50.892750	
300	SOUTH	NaN	UNDEVELOPED	-114.164420	50.986652	
...	...	...	...	...	...	...
300	POINT	(-114.16441971345019	50.986651529493514)			
309	POINT	(-114.02652092583433	50.909602044237715)			

[62 rows x 10 columns]  
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We were using community data points data set for sectors and community data type, so we needed to keep blank rows as we were getting accurate results without it, we created a csv file and then imported it to power bi keeping the blank rows.



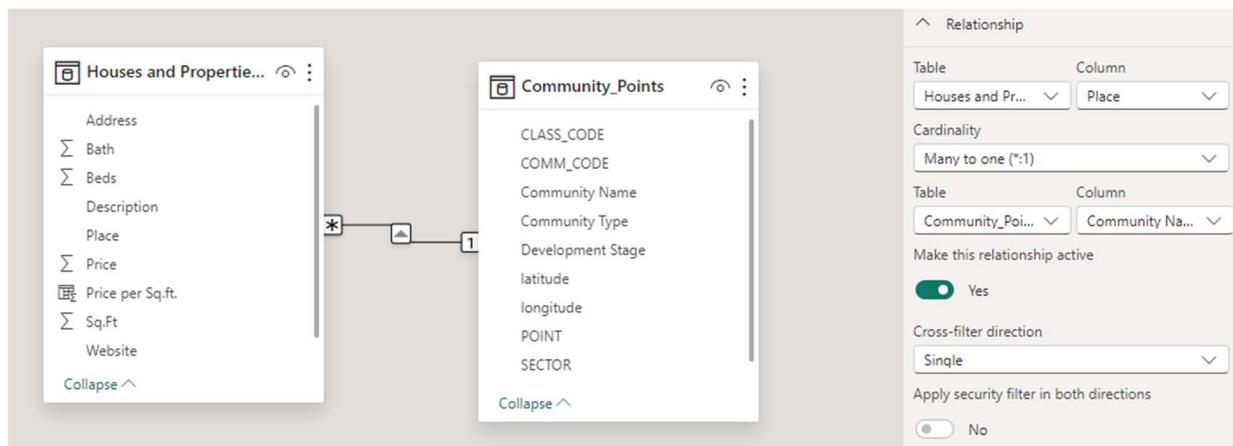
2. We checked Data types and made sure they are appropriate.

	A1 CLASS	I3 CLASS_CODE	A1 COMM_CODE	A1 NAME	A1 SECTOR	A1 SRG	A1 COMM_STRUCTURE	I2 longitude	I2 latitude
1	Residential	1	BED	BEDDINGTON HEIGHTS	NORTH	ESTABLISHED	1960s/1970s	-114.0850214	51.13163
2	Residential	1	EVN	EVANSTON	NORTH	COMPLETE	2010s	-114.1124526	51.17109
3	Residential	1	KIL	KILLARNEY/GLENGARRY	CENTRE	ESTABLISHED	1950s	-114.1317273	51.03154
4	Residential	1	BRA	BRAESIDE	SOUTH	ESTABLISHED	1960s/1970s	-114.1063659	50.95599
5	Residential	1	BLM	BELMONT	SOUTH	DEVELOPING	BUILDING OUT	-114.0552517	50.86868
6	Industrial	2	FHI	FOOTHILLS	EAST	N/A	EMPLOYMENT	-114.9820389	50.99396
7	Residential	1	RIC	RICHMOND	CENTRE	ESTABLISHED	1950s	-114.1184223	51.02971
8	Residential	1	CHN	CHINATOWN	CENTRE	ESTABLISHED	CENTRE CITY	-114.0626611	51.05088
9	Residential	1	CPF	COPPERFIELD	SOUTHEAST	DEVELOPING	BUILDING OUT	-113.9294735	50.91781
10	Industrial	2	EFV	EAST FAIRVIEW INDUSTRIAL	SOUTH	N/A	EMPLOYMENT	-114.0372414	50.9848
11	Residual Sub Area	4	13H	13H	SOUTH	N/A	UNDEVELOPED	-114.1128018	50.89234
12	Industrial	2	PEG	PEGASUS	NORTHEAST	N/A	EMPLOYMENT	-114.0064889	51.0944
13	Residential	1	ROY	ROYAL OAK	NORTHWEST	COMPLETE	2010s	-114.1197437	51.14266
14	Residential	1	CUR	CURRIE BARRACKS	WEST	ESTABLISHED	BUILDING OUT	-114.11248249	51.01571
15	Residential	1	HWS	HOTCHOSS	SOUTHEAST	DEVELOPING	BUILDING OUT	-113.8926547	50.91419
16	Residential	1	ASP	ASPEN WOODS	WEST	DEVELOPING	BUILDING OUT	-114.1083988	51.0462
17	Residual Sub Area	4	13N	13N	SOUTH	N/A	OTHER	-114.1437681	50.94484
18	Residential	2	HAY	HAYSBORO	SOUTH	ESTABLISHED	1950s	-114.0833403	50.97221
19	Residential	1	CHR	CHRISTIE PARK	WEST	ESTABLISHED	1980s/1990s	-114.1760167	51.04080
20	Residential	1	ALT	ALTADORE	CENTRE	ESTABLISHED	INNER CITY	-114.1018531	51.01595
21	Residential	1	UMR	UPPER MOUNT ROYAL	CENTRE	ESTABLISHED	INNER CITY	-114.0850185	51.02997
22	Residential	1	RED	RED CARPET	EAST	ESTABLISHED	1960s/1970s	-113.9427024	51.04019

3. We renamed the columns appropriately.

	Community Type	CLASS_CODE	COMM_CODE	Community Name	SECTOR	SRG	Development Stage	longitude	latitude
1	Residential	1	BED	BEDDINGTON HEIGHTS	NORTH	ESTABLISHED	1960s/1970s	-114.0850214	51.13163
2	Residential	1	EVN	EVANSTON	NORTH	COMPLETE	2010s	-114.1124526	51.17109
3	Residential	1	KIL	KILLARNEY/GLENGARRY	CENTRE	ESTABLISHED	1950s	-114.1317273	51.03154
4	Residential	1	BRA	BRAESIDE	SOUTH	ESTABLISHED	1960s/1970s	-114.1063659	50.95599
5	Residential	1	BLM	BELMONT	SOUTH	DEVELOPING	BUILDING OUT	-114.0552517	50.86868
6	Industrial	2	FHI	FOOTHILLS	EAST	N/A	EMPLOYMENT	-114.9820389	50.99396
7	Residential	1	RIC	RICHMOND	CENTRE	ESTABLISHED	1950s	-114.1184223	51.02971
8	Residential	1	CHN	CHINATOWN	CENTRE	ESTABLISHED	CENTRE CITY	-114.0626611	51.05088
9	Residential	1	CPF	COPPERFIELD	SOUTHEAST	DEVELOPING	BUILDING OUT	-113.9294735	50.91781
10	Industrial	2	EFV	EAST FAIRVIEW INDUSTRIAL	SOUTH	N/A	EMPLOYMENT	-114.0372414	50.9848
11	Residual Sub Area	4	13H	13H	SOUTH	N/A	UNDEVELOPED	-114.1128018	50.89234
12	Industrial	2	PEG	PEGASUS	NORTHEAST	N/A	EMPLOYMENT	-114.0064889	51.0944
13	Residential	1	ROY	ROYAL OAK	NORTHWEST	COMPLETE	2010s	-114.1197437	51.14266
14	Residential	1	CUR	CURRIE BARRACKS	WEST	ESTABLISHED	BUILDING OUT	-114.11248249	51.01571
15	Residential	1	HWS	HOTCHOSS	SOUTHEAST	DEVELOPING	BUILDING OUT	-113.8926547	50.91419
16	Residential	1	ASP	ASPEN WOODS	WEST	DEVELOPING	BUILDING OUT	-114.1083988	51.0462
17	Residual Sub Area	4	13N	13N	SOUTH	N/A	OTHER	-114.1437681	50.94484
18	Residential	2	HAY	HAYSBORO	SOUTH	ESTABLISHED	1950s	-114.0833403	50.97221
19	Residential	1	CHR	CHRISTIE PARK	WEST	ESTABLISHED	1980s/1990s	-114.1760167	51.04080
20	Residential	1	ALT	ALTADORE	CENTRE	ESTABLISHED	INNER CITY	-114.1018531	51.01595
21	Residential	1	UMR	UPPER MOUNT ROYAL	CENTRE	ESTABLISHED	INNER CITY	-114.0850185	51.02997
22	Residential	1	RED	RED CARPET	EAST	ESTABLISHED	1960s/1970s	-113.9427024	51.04019

4. We created Relationships between two data sets using Place and Community Name.



We connected the "Place" column from the "Housing Price & Real Estate - 2023" dataset to the "Community Name" column in the "Community Points" dataset. This link creates a many-to-one relationship, suggesting that several properties listed in the "Housing Price & Real Estate - 2023" dataset could match one community in the "Community Points" dataset.

## 10. Evaluation:

After reviewing the entire process once more, let's now outline the next steps. With all the necessary data in hand, we are ready to formulate business questions. Subsequently, we'll develop visualizations to address these questions and deploy them in Power BI. Finally, we'll present our analysis through a PowerPoint presentation.

## 11. Business Questions

### 1. What is the average price per square foot for homes in different communities, and how does it vary across sectors? And how does the community type (Industrial and Residential) influence the price of homes for sale?

- **Average Price per Square Foot for Homes in Different Communities and Sectors:**  
The below dashboard includes a table and a bar chart that display the average price per square foot for various communities across different sectors (CENTRE, EAST, NORTH, NORTHEAST, NORTHWEST, SOUTH, SOUTHEAST, WEST)

- **Variation Across Sectors:**

The bar chart visually represents these differences, showing a wide range of average prices per square foot within sectors.

- **Variation Across Community Type:**

It shows if it's a residential community or industrial based on Average of Price.

- **Count of Properties:**

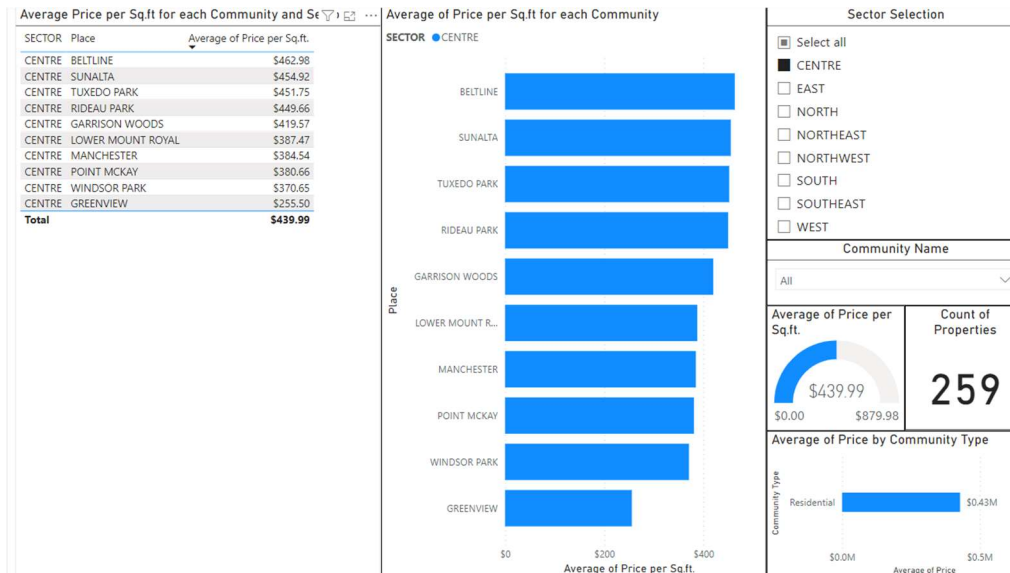
It shows the number of properties listed in that community.

By this dashboard will analyze and answer the above business question

### Top 10 and bottom 10 in Centre sector.

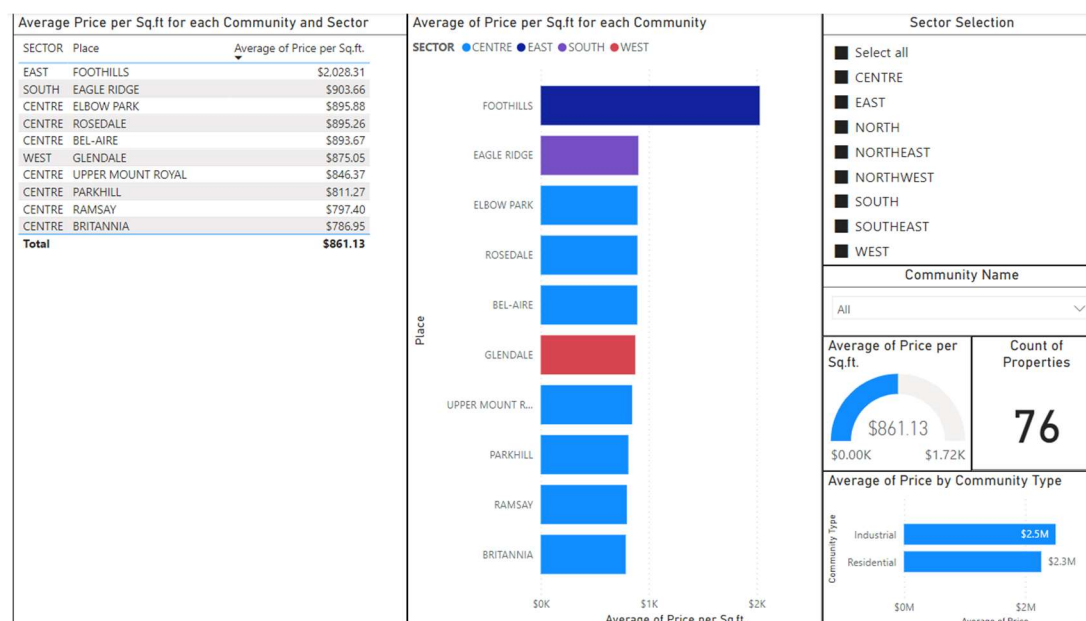


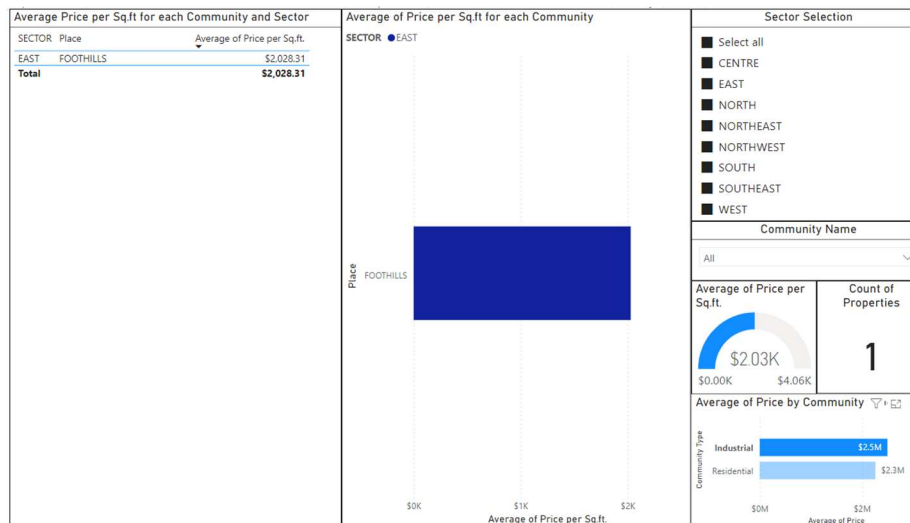
In the above dashboard image we can see the top 10 in the CENTRE sector, Elbow Park has an average price of \$895.88 per square foot which is one of the highest in that sector and the average price per sq ft in that Sector for that top 10 is around \$792.81 which is residential class and total number of properties listed is 120 and the average of property price is \$1.9M.



In the above dashboard image, we can see the bottom 10 in the CENTRE sector, Greenview has an average price of \$255.50 per square foot which is one of the lowest in that sector and the average price per sq ft in that sector for that bottom 10 is around \$439.99 which is residential class and total number of properties listed is 259 and the average of property price is \$0.43M.

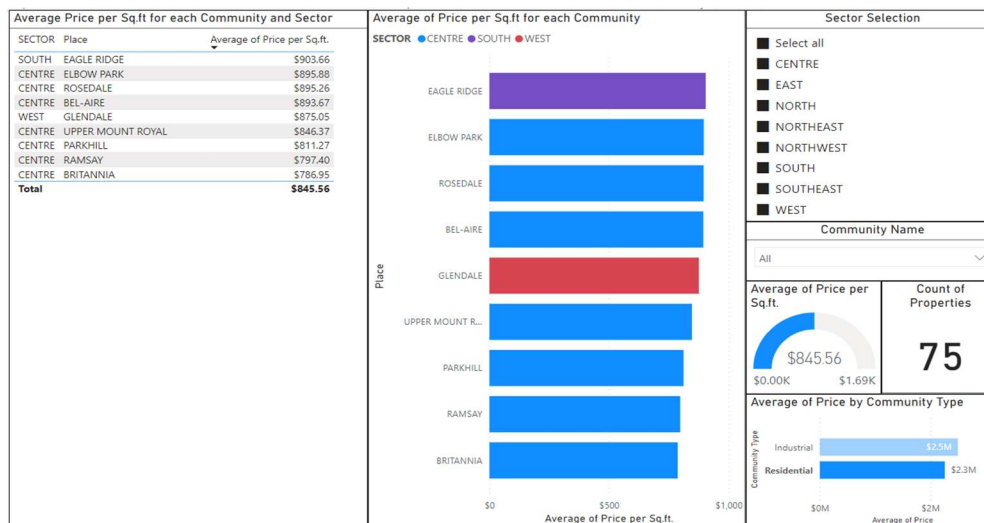
## For Top 10 across all sectors





In the above dashboard images if we consider all the sectors instead of just center, we can see the top 10 across all sectors, Foothills Community has an average price of \$2028.31 per square foot which is one of the highest across all sectors which is in East sector, and it is an industrial class that's the reason for high price. the average price per sq ft for that top 10 is around \$861.13 which comes under both residential and industrial class and total number of properties listed is 10 and the average of property price is \$2.3M if it's a residential class and \$2.5M in industrial class. Also, we can see that the north, northeast, northwest and southeast are not available in the top 10.

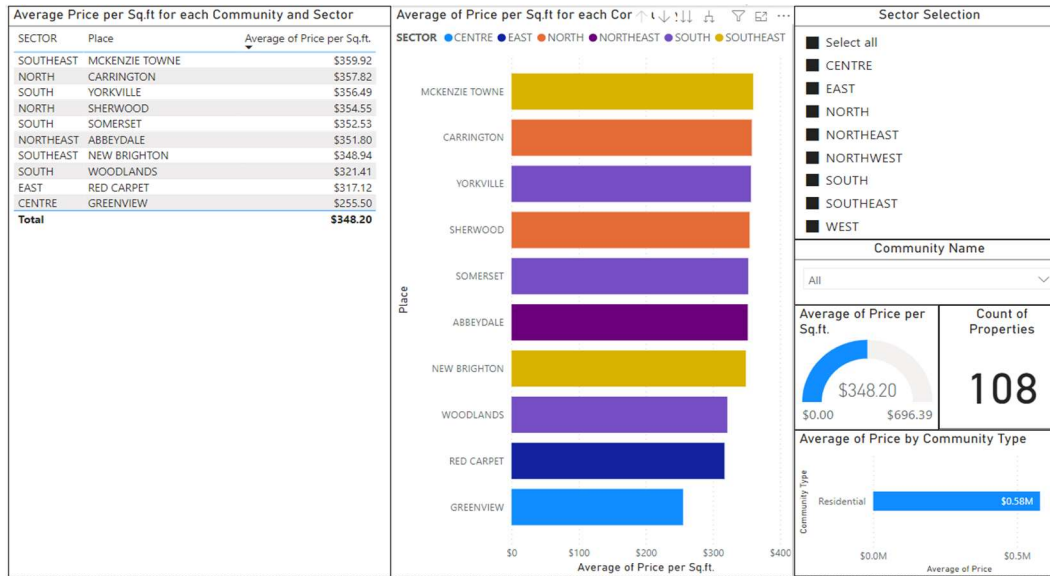
### For Top 10 across all sectors in Residential Community type



In the above dashboard images if we consider all the sectors instead of just center and ignore the industrial class which is foothill community and only consider residential class, we can see the top 10 across all sectors for residential class, Eagle Ridge has an average price of \$903.66 per square foot which is one of the highest across all sector in residential class which is in South sector and the average price per sq ft for that top 10 is around \$845.56 which comes under residential and the average of property price is \$2.3M.

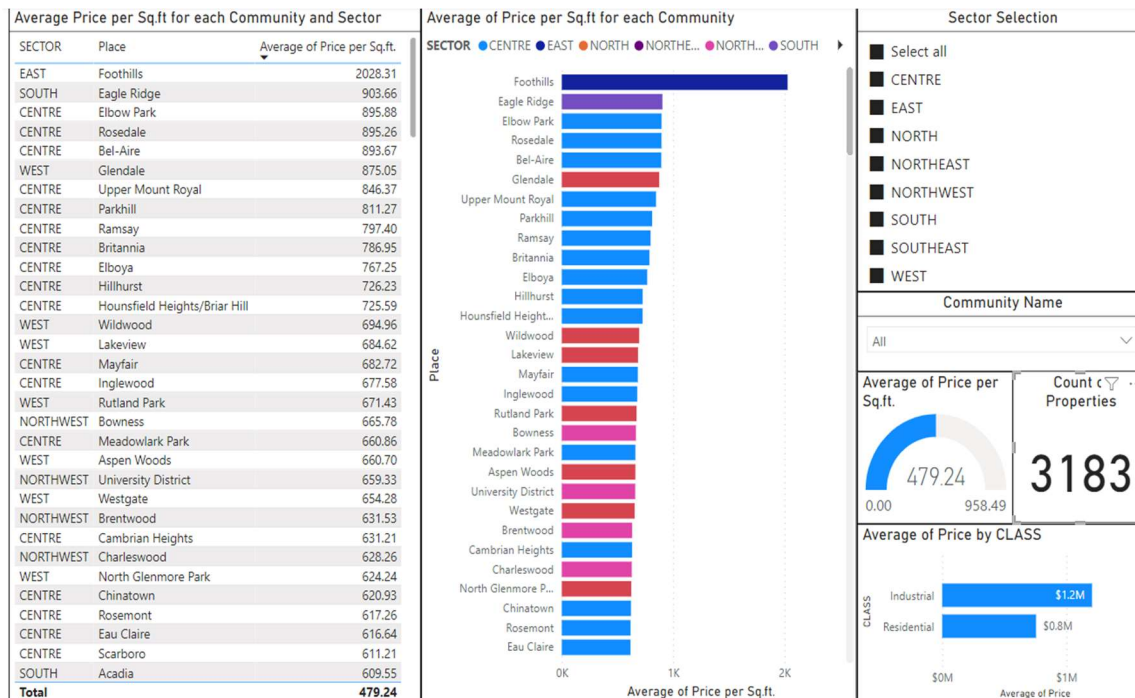


## For Bottom 10 across all sectors



In the above dashboard image if we consider all the sectors instead of just center, we can see the bottom 10 across all sectors, Again Greenview has an average price of \$255.50 per square foot which is one of the lowest across all sector even though it's in Centre sector and the average price per sq ft for that bottom 10 is around \$348.20 which is residential class and total number of properties listed is 10 and the average of property price is \$0.58M. Also, we can see that the west and northwest sector is not available in the bottom 10.

We can see by the below dashboard without any filters:



Industrial properties have a higher average price at \$1.2 million, whereas residential properties have a lower average price at \$0.8 million. This suggests that industrial properties tend to be priced higher per square foot compared to residential properties.

Overall, the dashboard provides a comprehensive view of how home prices per square foot vary across different communities and sectors, and how community type affects these prices. By using this buyer can decide where they want to purchase the kind of property they are looking for and the average price per square foot and total properties available to look at in that community, etc.

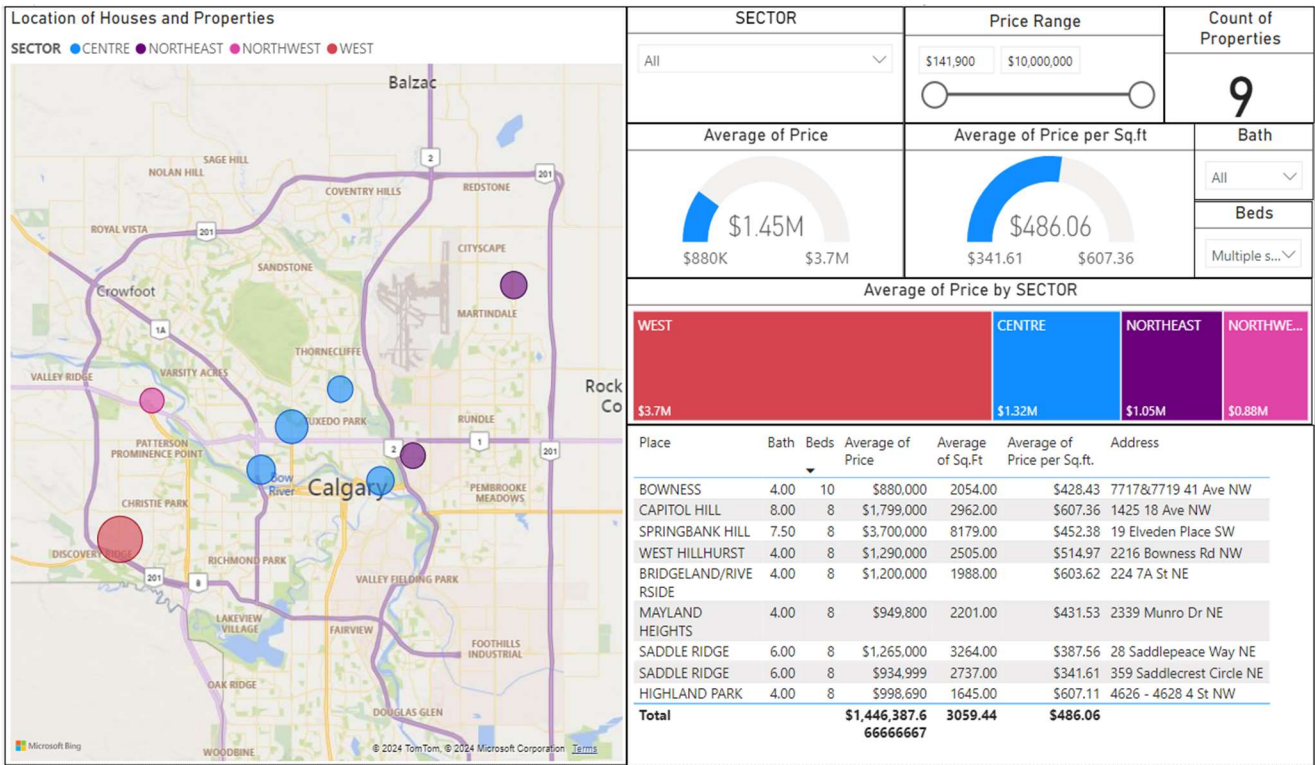
This analysis can be crucial for strategic decisions related to investment or development in specific sectors.

2. In pursuit of investment opportunities, what type of property would be optimal in terms of maximizing the number of bedrooms? Additionally, in which community is such a property typically located, and what is the associated selling price?

If we look at properties with the maximum number of bedrooms listed, with multiple selection from 10 and 8 bedrooms with all number of bathrooms with no price range limit.

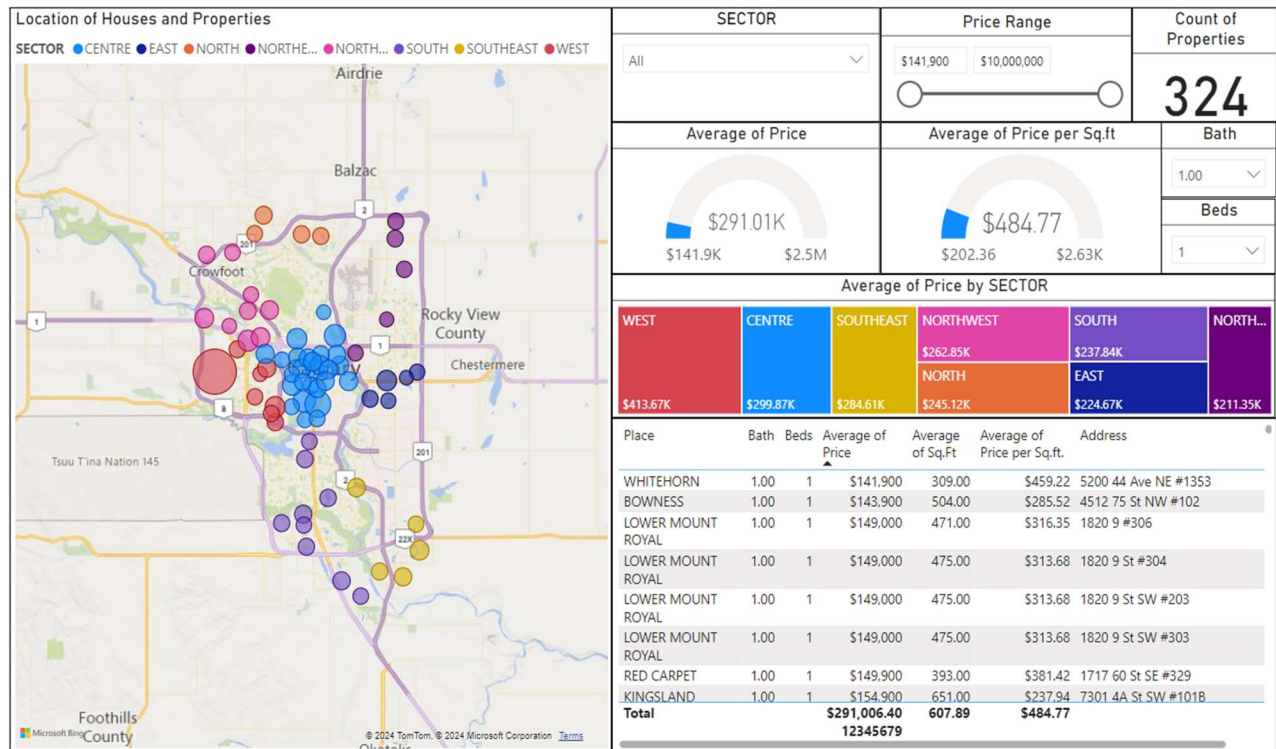
We can see from the below dashboard that there are 9 such properties which are suitable for investment with max of bedrooms in Centre, Northeast, Northwest, and west sectors.

Using this dashboard, stake holders can compare it with other properties which are similar and invest in these properties and list it for rent so that they can get good rental income.





### 3. For a single individual with the financial means to purchase a property featuring one bedroom and one bathroom, what would be the anticipated cost of such a unit?



Based on the above dashboard we can see that a single individual who is looking for a 1 bed 1 bath property they have total of 324 properties listed based on the price range 141,900 \$ to 2,50,0000 \$ and 307 sq. ft. to 1057 sq. ft. to choose from.

With this information, stakeholders can make selections based on sq. ft., community, and price to determine the most suitable option.

## 12. Deployment:

As Mentioned in Evaluation we deployed all the Power Bi visuals in Power Bi Server and Presentation will and other related Files are uploaded in Brightspace which can be accessed by the stakeholders to analyze and decide on their investment.

Link for Power Bi Server : [Final assignment - NW - Power BI](#)

<https://app.powerbi.com/groups/me/reports/84bd6331-498c-42da-9409-870add279e4a/ReportSection?experience=power-bi>

## 13. Conclusion:

Through our comprehensive analysis of the Canadian housing market in 2023, utilizing datasets from Kaggle and Calgary Open Data, we have gained valuable insights into housing price trends, regional variations, and the impact of community types on property values. We employed the CRISP-DM methodology, ensuring a structured approach from data understanding to visualization and analysis.

### Key Findings:

1. **Regional and Sectoral Variations:** Our analysis revealed significant disparities in housing prices across different regions and sectors. From affluent neighborhoods commanding high prices per square foot to more affordable areas, buyers and investors have a wide range of options to consider.
2. **Influence of Community Type:** We observed that community type, whether residential or industrial, plays a crucial role in determining property values. Industrial properties tended to have higher average prices, highlighting the importance of understanding the specific dynamics of each community.
3. **Optimal Property Types for Investment:** Our exploration of properties with the maximum number of bedrooms indicated potential investment opportunities in various sectors. Understanding the demand for specific property types can guide investors in making informed decisions.
4. **Affordability and Accessibility:** For individuals seeking properties with one bedroom and one bathroom, our analysis identified a range of options across different price points and sizes. This information can empower prospective buyers to navigate the market efficiently.

### 13. Recommendations:

1. **Diversification of Investment Portfolio:** Investors should consider diversifying their portfolios across different regions and property types to mitigate risks and capitalize on emerging opportunities. Understanding market dynamics and demographic trends is essential for strategic decision-making.
2. **Community-Level Analysis:** Further research into community-level factors, such as infrastructure development, amenities, and future growth projections, can provide deeper insights into long-term investment prospects. Collaborating with local stakeholders and real estate professionals can enhance understanding and decision-making.
3. **Continuous Monitoring and Adaptation:** The real estate market is dynamic, influenced by economic shifts, regulatory changes, and societal trends. Regular monitoring of market indicators and updating analytical models will ensure relevance and accuracy in decision support.
4. **Investment in Data Analytics:** Given the importance of data-driven insights in the real estate sector, investing in advanced analytics tools and expertise can provide a competitive edge. Leveraging predictive modeling and machine learning algorithms can enhance forecasting capabilities and risk assessment.
5. **Improvement to Dataset:** we suggest creating a dataset based on postal codes with sector data for Calgary along with nearby hotspots or malls. This will provide more accurate and localized insights into housing market trends. Collaborating with local authorities can ensure data accuracy. Integrating this dataset will help stakeholders make informed decisions tailored to specific neighborhoods and sectors within Calgary, enhancing investment strategies.

---Thank You---