



# Unlocking Shared Housing Supply in Canada: Happipad Final Presentation

Yuzhu Han, Foster Lockerbie, Jingran Zhao & Litao Zheng



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# Our Team



**Yuzhu Han**

Background in  
statistics and urban  
planning



**Foster Lockerbie**

Background in  
psychology  
(statistics focus)



**Jingran Zhao**

Background in  
information  
management



**Litao Zheng**

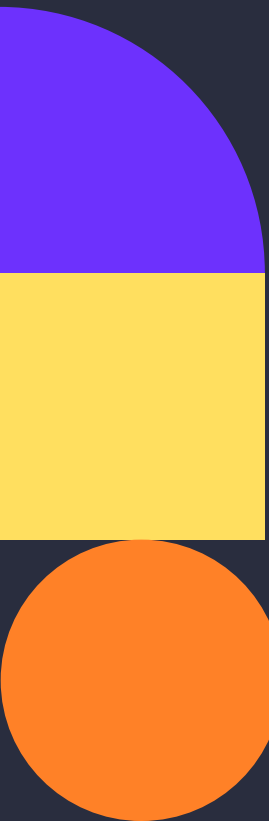
Background in  
Mathematics and  
Statistics

# What is Happipad?

Housing platform geared toward “unlocking underutilized housing”. Focused on providing affordable homes for seniors, refugees, and students.

Matches landlords with renters based on comprehensive profiles with list of habits and preferences.

# Research questions



**Can we effectively represent the housing demand and trends through dashboarding?**

**Can we predict which home listings will be rented?**

**Can we predict the appropriate price for rental listings?**

**Can we use any large language models to improve our predictive modelling?**

# Deliverables

## Dashboards

Property and renter  
monitoring:  
Property overview  
Renter overview

Two Versions:  
Dash and Tableau

## Predictive Modelling

Predicting price of  
rental spaces based  
on home features

Using NLP to inform  
price predictions based  
on text responses

## Final Report

Formal project report  
including:

Project research  
questions

Methods

Predictive modelling  
results and analysis

Dashboard explanation

# Data Overview

## Renters Dataset

- Desired lease term
- Budget
- City
- ...

## Hosts Dataset

- Location
- Preferred gender of roommates
- Disabilities
- ...

## Properties Dataset

- Rent price
- Property locations
- Furnishing details
- ...

## Contracts Dataset

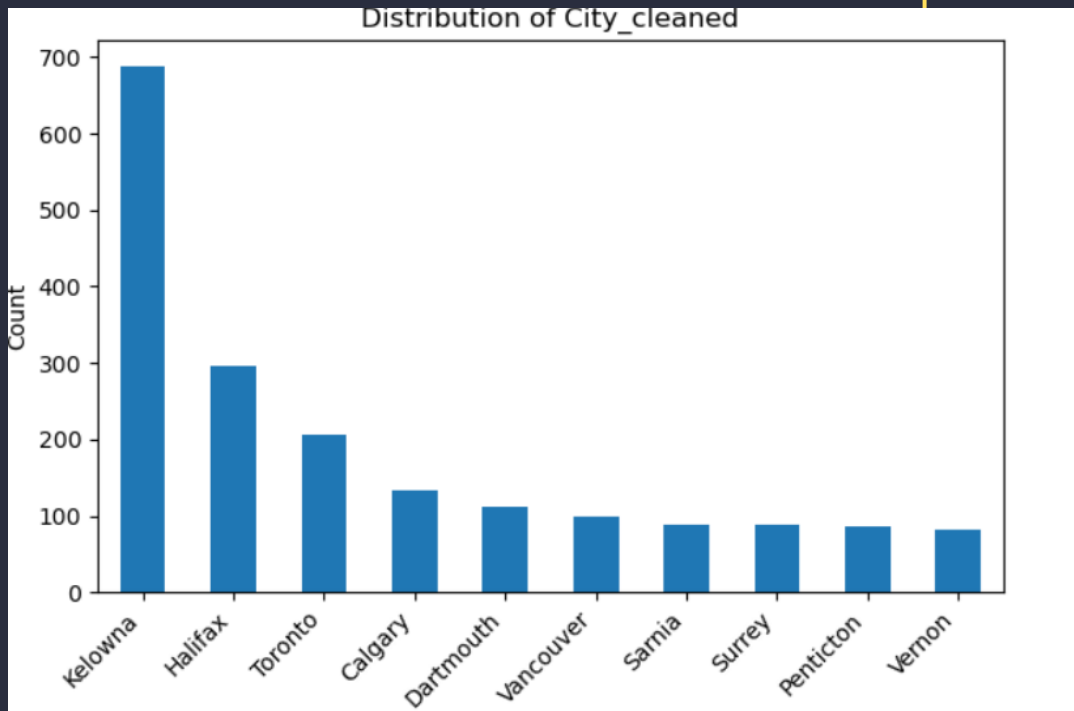
- Monthly price of rent
- Lease term
- Location
- ...

## ■ Data Cleaning

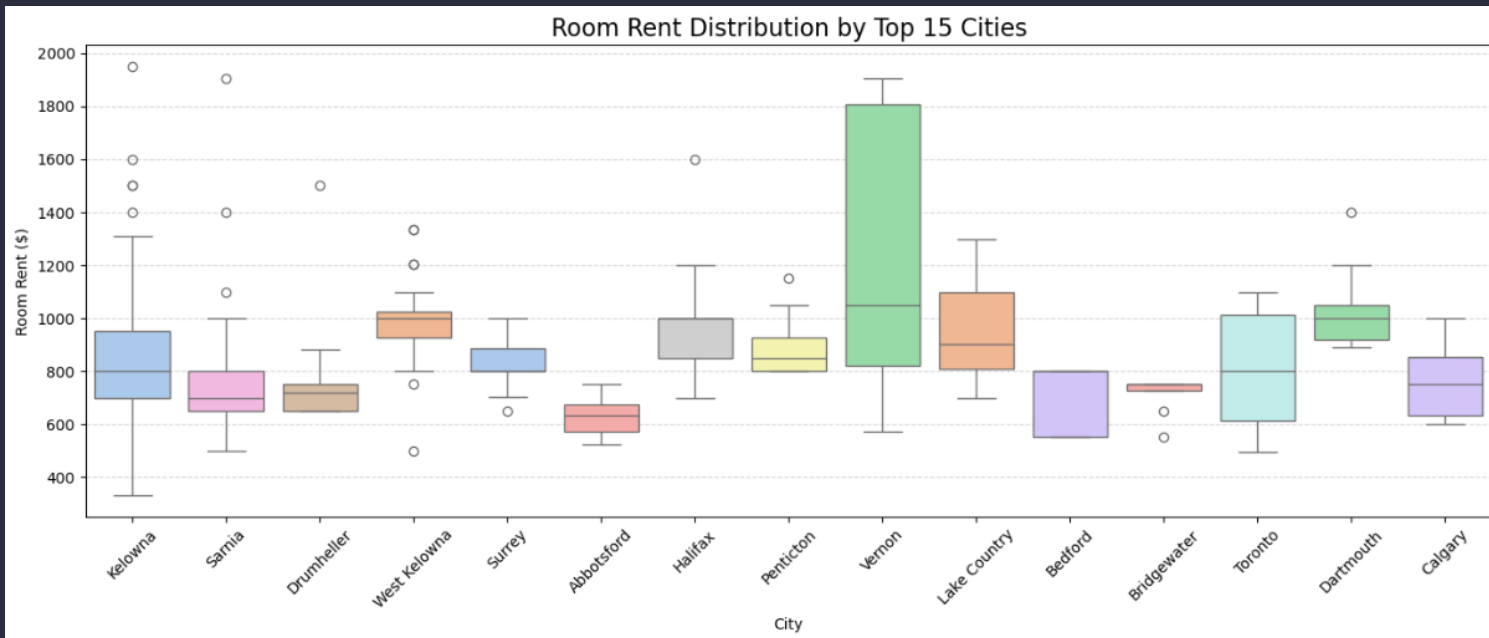
- Investigate outliers
  - determine which values are considered outliers and if they should remain in the dataset
- Remove test properties and listing
- Standardize columns (postal code, address, city, province)
- Binarize columns that contain multiple values
- Convert data types



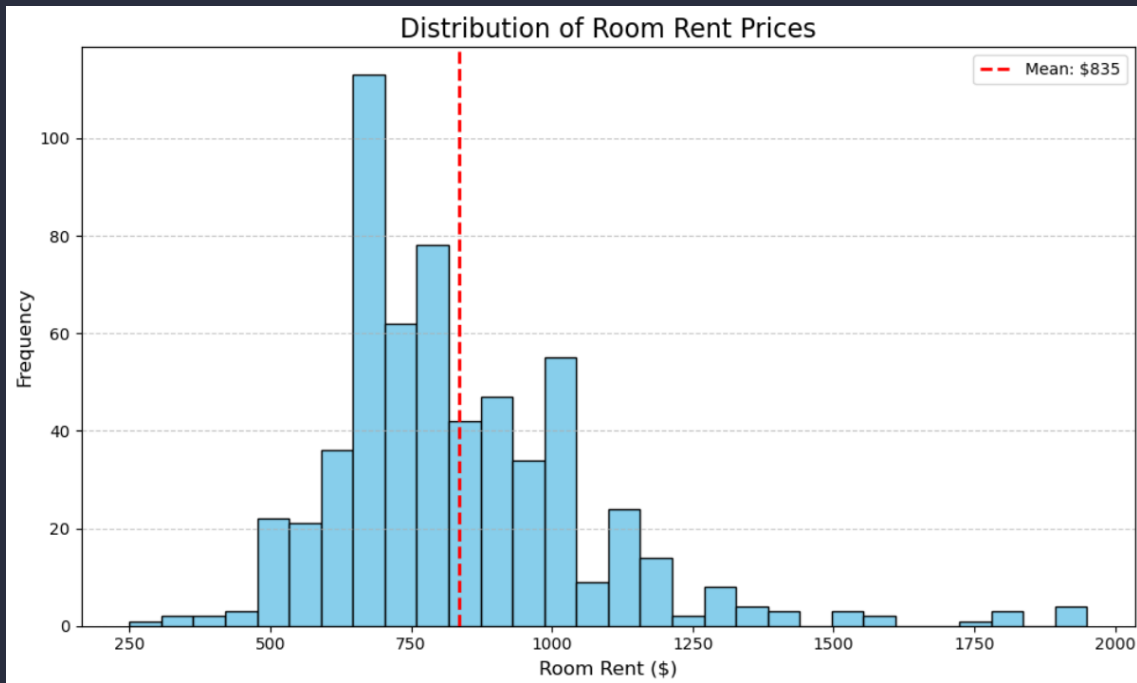
## EDA - Hosts



# EDA - Contracts

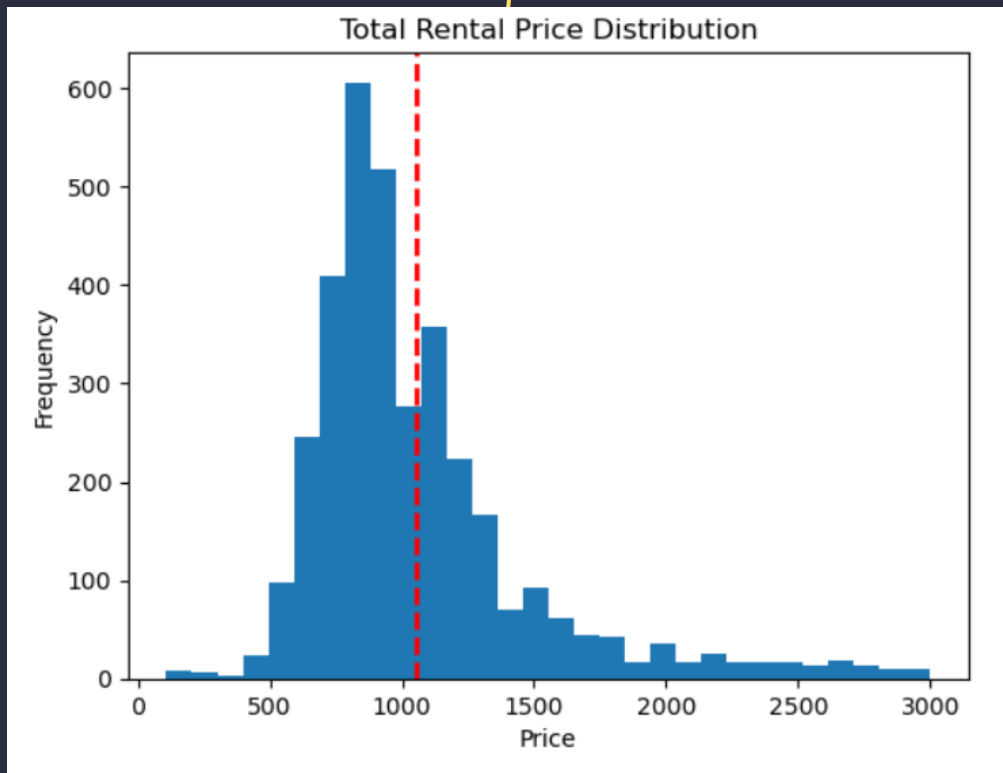


# EDA - Contracts

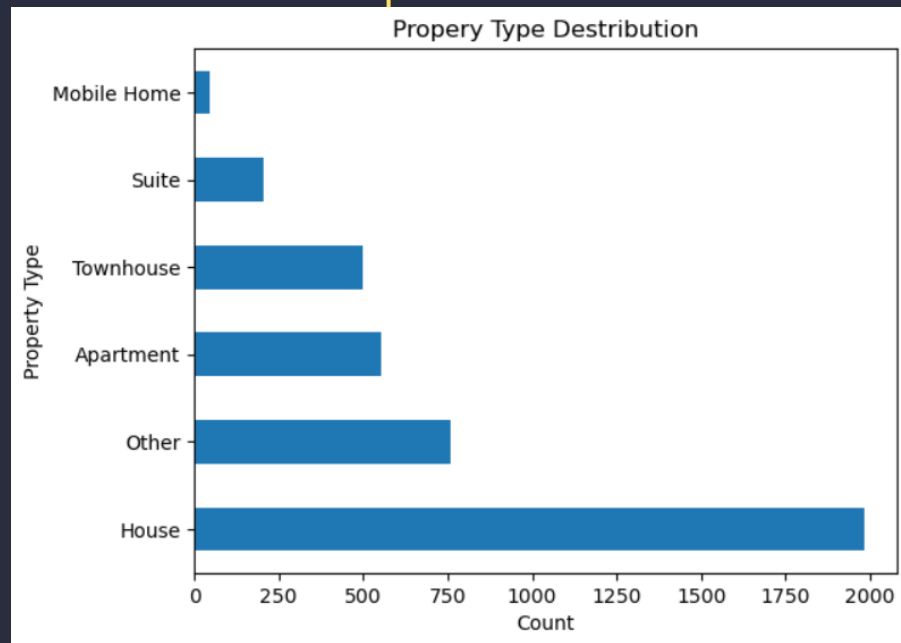
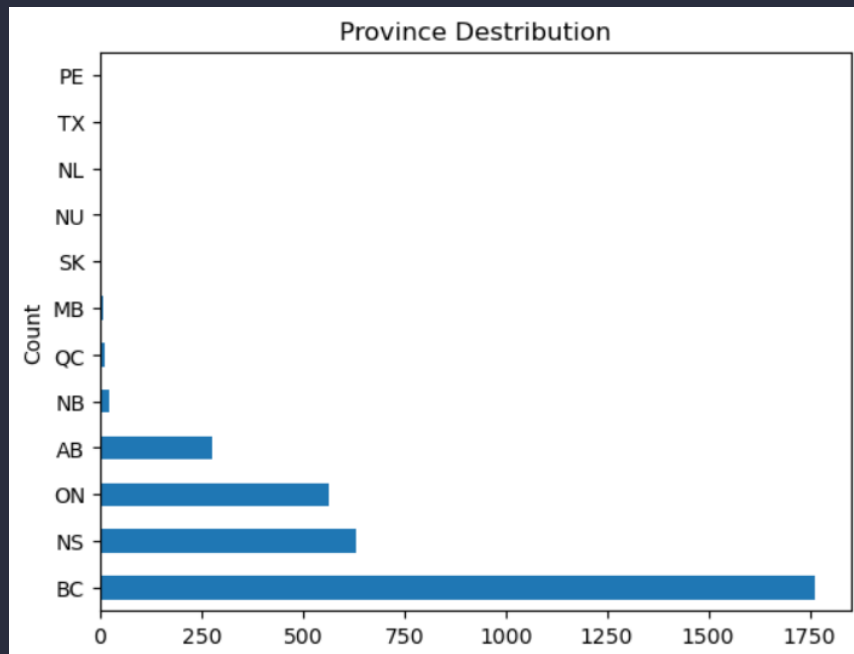


# EDA - Properties

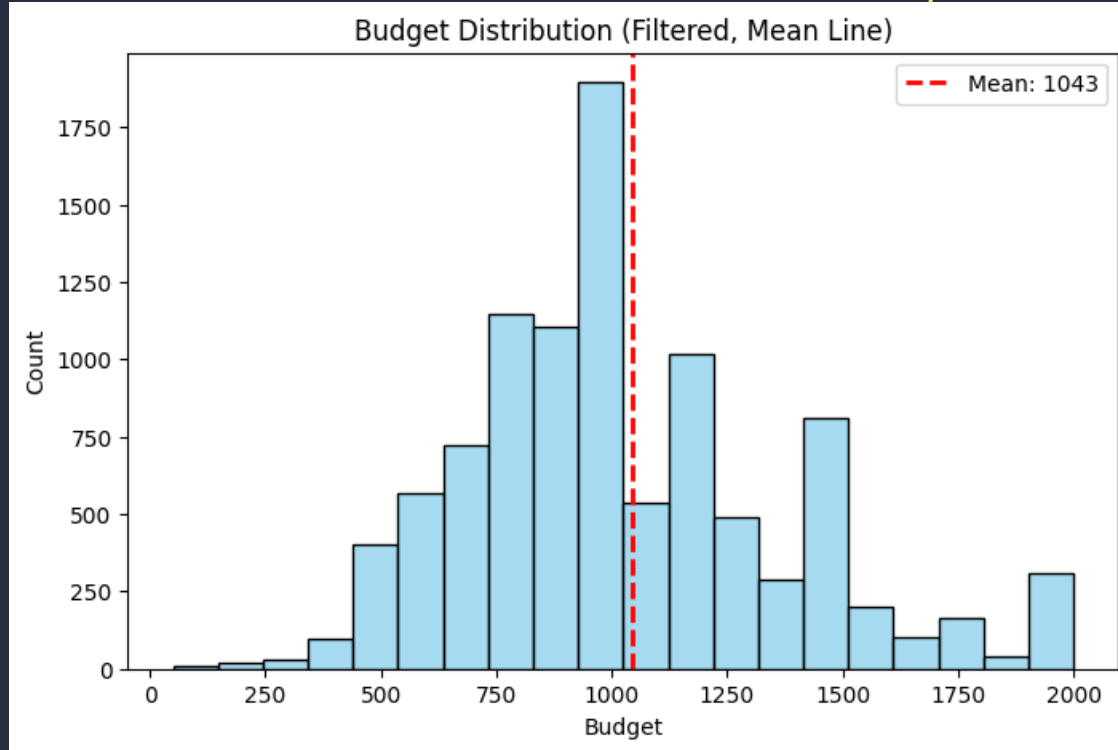
```
count    3666.000000
mean     1098.124326
std      2091.288179
min       0.000000
25%      756.000000
50%      947.040000
75%      1188.000000
max     118800.000000
Name: Total, dtype: float64
```



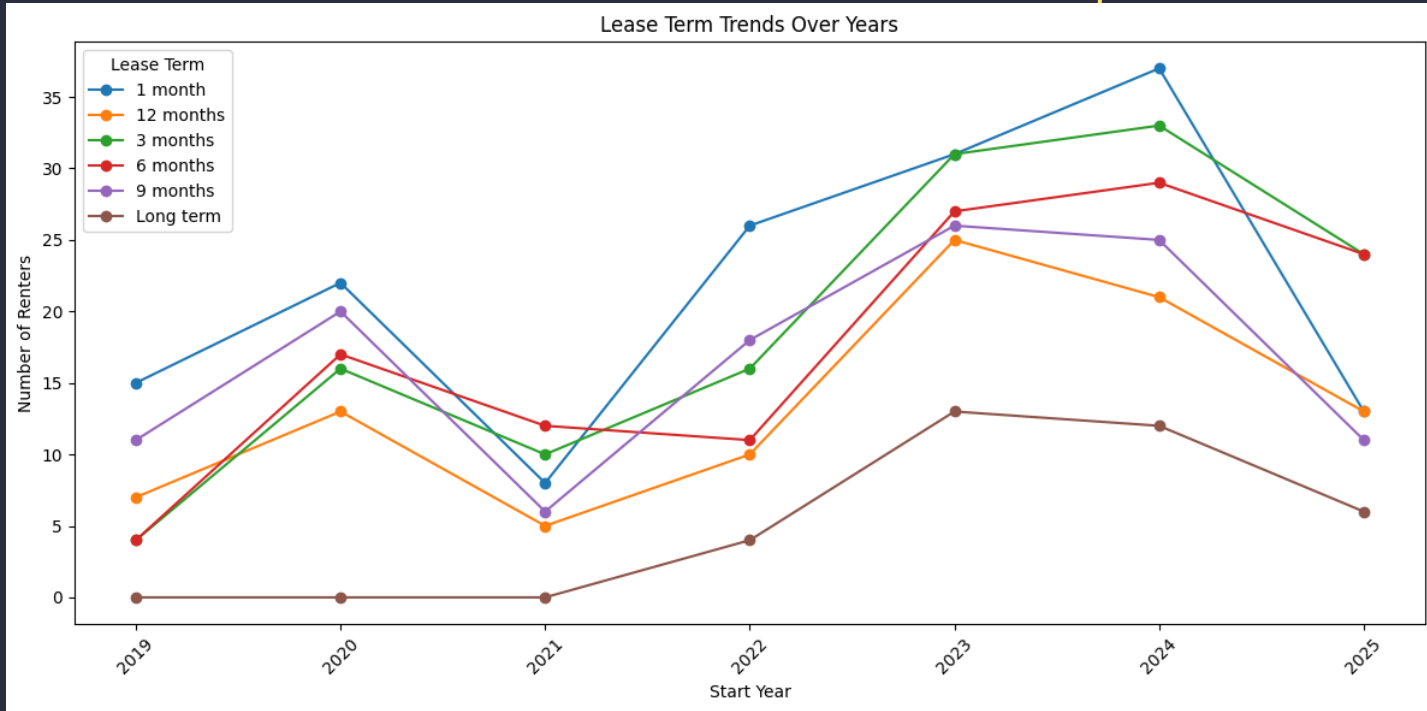
# EDA - Properties



# EDA - Renters



# EDA - Renters



# Results:

## Predictive Modelling - price predictions

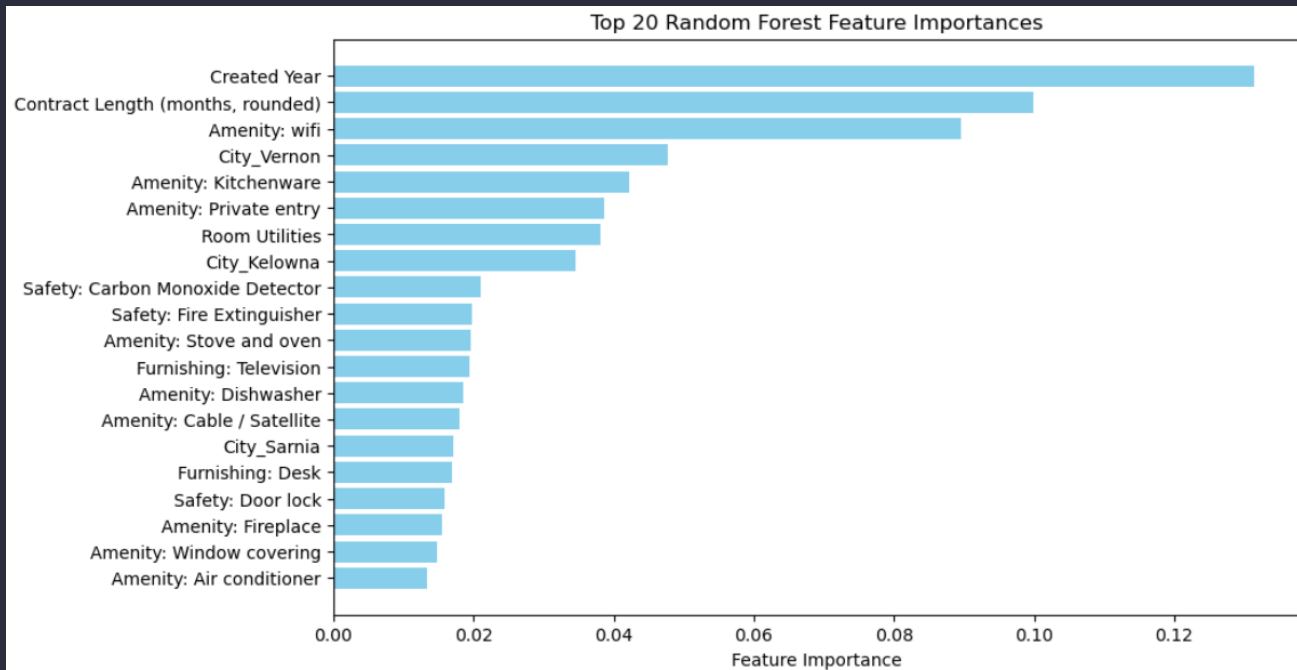
Model Name	RMSE	Relative Error
XGBoost (target encoding)	137.41	16.46%
XGBoost (one-hot encoding)	141.89	16.99%
Random Forest	133.60	16.00%
LightGBM	137.42	16.46%
Random Forest + NLP	140.43	16.82%



# Results:

## Predictive Modelling

### - price predictions



## Results:

# Predictive Modelling - Rental Likelihood Classifier

- Target variable: a binary column indicating whether the property has a rental record in the contracts dataset
- 5-fold Random Forest:

```
Fold 1 – Accuracy: 0.921, F1: 0.179  
Fold 2 – Accuracy: 0.926, F1: 0.062  
Fold 3 – Accuracy: 0.925, F1: 0.141  
Fold 4 – Accuracy: 0.906, F1: 0.136  
Fold 5 – Accuracy: 0.921, F1: 0.179  
Mean Accuracy: 0.920  
Mean Precision: 0.902  
Mean Recall: 0.077  
Mean F1 Score: 0.140
```

		Predict	
		0	1
True	0	3693	4
	1	321	27

# ▀ NLP - Ollama

Ollama is a command-line platform for running large language models (LLMs) on local computer for offline text generation and analysis

- natural language processing tasks such as summarization, text classification, etc.

## **Mistral:**

- Good balance between accuracy and speed
- Run efficiently on local hardware

# ▀ NLP - Ollama

## Expected output columns:

- `number_of_people`: number of people mentioned in the description
- `bedrooms`: how many bedrooms for rent
- `pets_allowed`: whether pets are allowed on the property
- `property_size`: the size of the property
- `shared_spaces`: a list of shared spaces in the rental listing
- `bathroom_type`: the bathroom would be shared or private
- `nearby_amenities`: a list of nearby amenities, such as bus stop
- `unique_features`: unique features of the property

# ■ NLP - Ollama

## Input prompt 1:

*Return a JSON object with:*

- *Number Of People (how many people the space is for or referenced)*
- *Number Of Pets (how many pets are mentioned)*
- *Property Size (e.g., small, medium, large, unknown)*
- *Shared Spaces (e.g., kitchen, living room)*
- *Bathroom Type (shared or private)*
- *Nearby Facilities (e.g., bus stop, store)*

*If the information is missing, return "unknown".*

# ▀ NLP - Ollama

Example output 1:

Number Of People	Number Of Pets	Property Size	Space Type	Shared Spaces	Bathroom Type
Unknown	Unknown	Unknown	Room	['Kitchen', 'Bathroom']	Shared
2	unknown	unknown	apartment	['living room', 'kitchen']	private
1	2	Unknown	Room	['kitchen', 'bathroom', 'living room', 'laundry	Shared
Unknown	Unknown	Unknown	Room	['kitchen', 'living room']	Unknown
unknown	unknown	unknown	unknown	unknown	unknown
unknown	unknown	unknown	unknown	unknown	unknown
unknown	unknown	unknown	master bedroom	['kitchen', 'laundry room', 'cozy living room']	private

# ▀ NLP - Ollama

Input prompt final:

Schema:

```
{(  
  "number_of_people": "integer or 'unknown'",  
  "bedrooms": "integer or 'unknown'",  
  "pets_allowed": "true, false, or 'unknown'",  
  "property_size": "'small', 'medium', 'large', or 'unknown'",  
  "shared_spaces": "comma-separated string or 'unknown'",  
  "bathroom_type": "'private', 'shared', or 'unknown'",  
  "nearby_amenities": "comma-separated string from [bus, store, recreation  
centre/pool, school and university] or 'unknown'",  
  "unique_features": "semicolon-separated string or 'unknown'"  
})
```

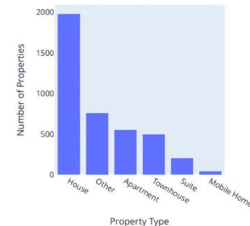
# ▀ NLP - Ollama

Example output final:

number_of_people	bedrooms	pets_allowed	property_size	shared_spaces	bathroom_type	nearby_amenities	unique_features	
unknown	1	unknown	unknown	unknown	private	bus, unknown	parking	
unknown	unknown	unknown	medium	unknown	unknown	bus, store, recreation centre/pool,	five minute walk to the beach;	
unknown	unknown	unknown	unknown	living room	unknown	bus, store, recreation centre/pool,	none	
unknown	unknown	unknown	unknown	unknown	unknown	unknown	suitable for students	
unknown	1	unknown	unknown	living room, kitchen	unknown	bus, store, recreation centre/pool	fireplace	
unknown	1	unknown	unknown	unknown	unknown	unknown	unknown	
unknown	1	unknown	unknown	unknown	unknown	unknown	Queen bedroom	
1	unknown	unknown	medium	kitchen	unknown	bus, store, recreation centre/pool	furnished room, electricity, W	
unknown	unknown	unknown	unknown	unknown	private	unknown	unknown	
unknown	1	unknown	medium	unknown	unknown	bus, school (UBCO)	spacious	



- Property Status
- Geographical Distribution
- Property Characteristics



# Dashboard(Dash): Property Overview

## Property and Renter Monitoring

### Property Overview

### Renter Overview

#### Active

Count: 81

Percentage: 12.29%

#### Signed

Count (2025-05): 4

+0.00%

#### Avg. Price

Latest (2025-05): \$829.17

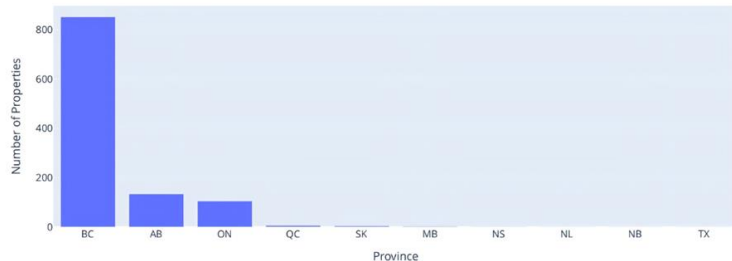
-10.90%

## Property Status

### Year Range Filter

2019 2020 2021 2022 2023 2024

### Number of Properties by Province



### Overall Property Count by City



## Location Distribution

# Dashboard(Dash): Property Overview

## Other Features

Filter by Province:

Select Province(s)

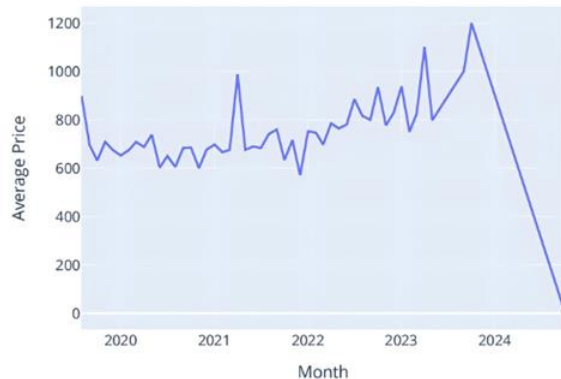
Select Year(s):

Select Year(s)

Filter by City:

Select City(ies)

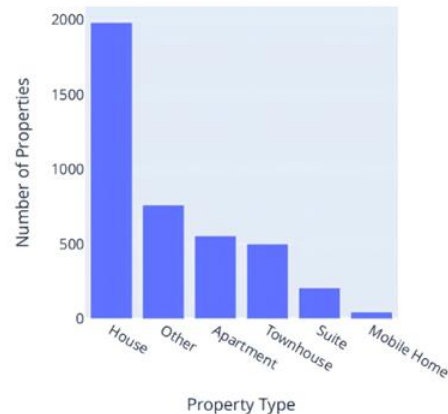
Overall Average Monthly Price



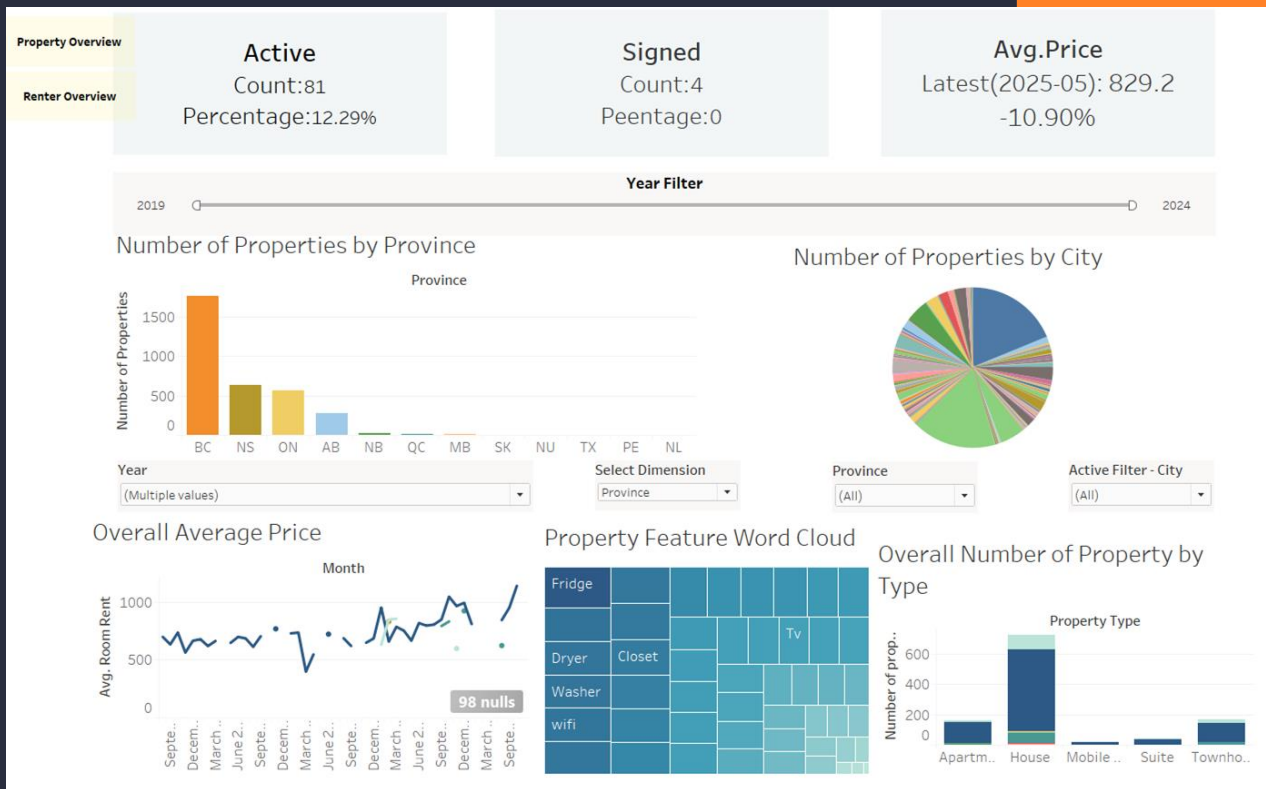
Property Feature Word Cloud



Overall Number of Properties by Type

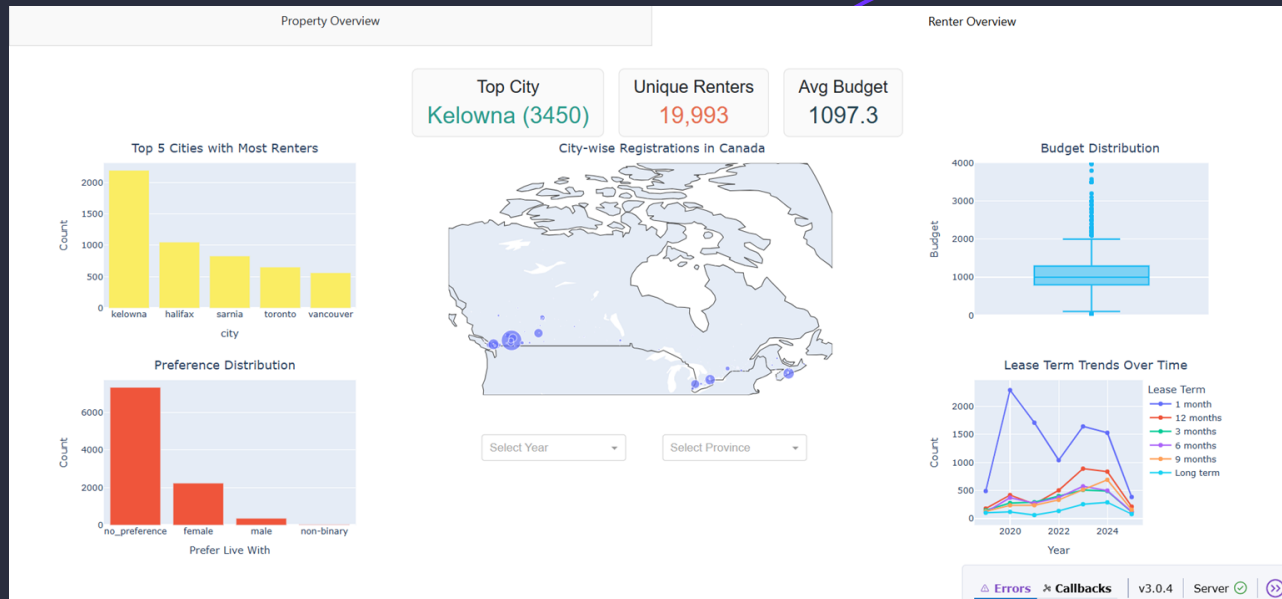


# Dashboard(Tableau): Property Overview

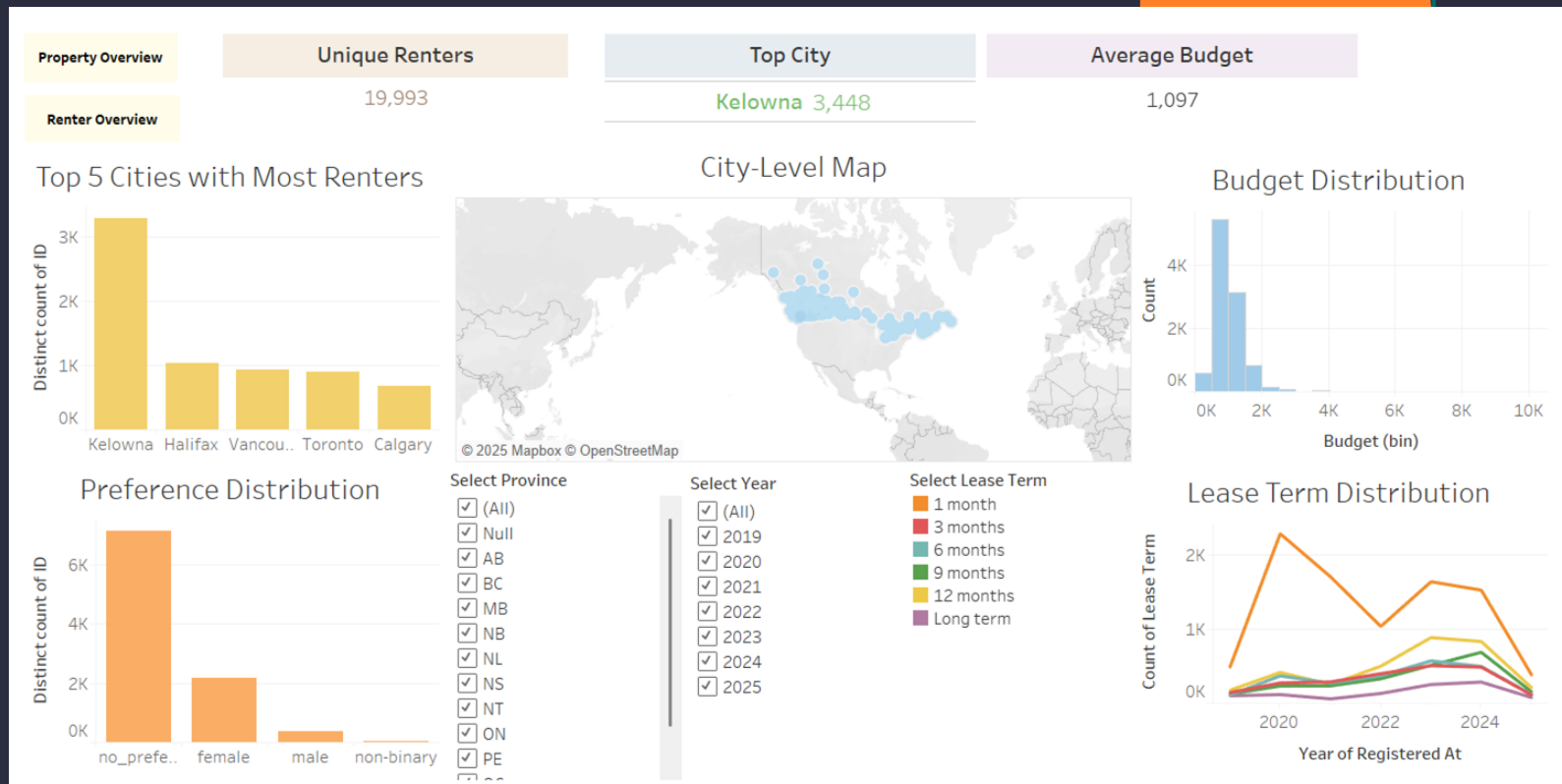


# Dashboard(Dash): Renter Overview

- Basic Information card: Top City, Unique Renters, Average Budget
- Canada map showing renter distribution
- Budget Distribution
- Lease term distribution
- Filter dropdowns: Year, Province



# Dashboard(Dash): Renter Overview



## Conclusion

- Two dashboards (Property and Renter Overview):
  - Help Happipad monitor business
  - Help Happipad understand their clientele
- Modelling Framework
- Ollama with Mistral

### Future works:

- Monitor model performance as dataset grows
- Build a streamlined process to continuously update the dashboard



# Acknowledgement

Happipad

MDS Capstone

Dr. Scott Fazackerley



**Thank you!**