## The Battle of Students

A data analysis project for apartment-seeking students in Graz, Austria.

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## Introduction

## **Background**

#### Graz

- 300,000 residents
- 60,000 students
- Strong competition in apartment-hunting

#### Renting vs. Buying

- Many companies in and around Graz
- Students often move permanently after their studies
- Low interest rates

#### **Problem**

## Renting

- Current situation in a borough of interest
- Cheap rent, close proximity to university

### **Buying**

- Where ist it currently affordable?
- Will value increase/stay/decrease?

#### **General**

- Information source?
- Relying on assumptions and recommendations

## **Target Audience**

#### **Target Audience**

- Students planning to study in Graz
- Looking for an apartment to rent
- Looking for property to buy

## **Data**

#### **Data Sources**

#### **Graz**

https://de.wikipedia.org/wiki/Liste\_der\_ Stadtbezirke\_von\_Graz

#### **Rent and Realty Prices**

- https://www.immobilienscout24.at/
- https://www.immowert123.at/

#### **Foursquare**

Vibrancy of borough based on availability of venues

#### **Location Data**

• geopy: Nominatim, ArcGIS

## **Data Acquisition**

#### **Pandas**

- For data in tabular form
- Wikipedia
- https://www.immobilienscout24.at/

#### **Beautiful Soup**

- For less nicely provided data
- https://www.immowert123.at/

## Data Cleaning, Data Preparation, Feature Selection

#### Graz

- Borough names
- Area
- Population Density

## https://www.immobilienscout24.at/

Use entire data

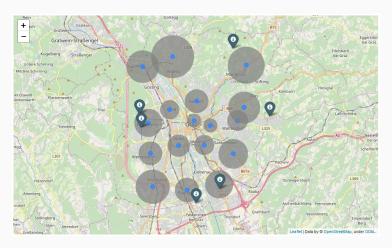
## https://www.immowert123.at/

• Filter for entries with "wohnung" (apartment) in it

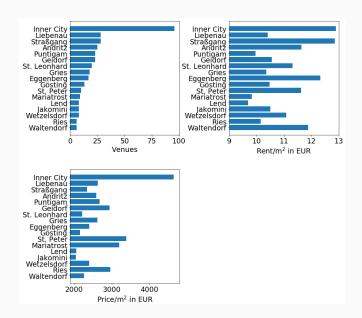
# Methodology

## **Starting Point**

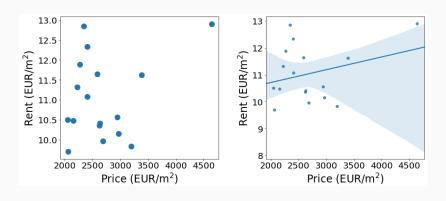
## Boroughs, Districts, Area Size



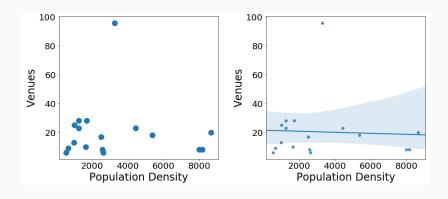
## Exploratory Data Analysis: Vibrancy, Rent, Realty Price



## Exploratory Data Analysis: Realty Price vs Rent



## **Exploratory Data Analysis: Population Density vs Vibrancy**



## **Exploratory Data Analysis**

## **Conclusions from Data Analysis**

- Variables are independent
- Small data set but maximum information
- Rent and prices may lead to different clusters

## Clustering

#### **Important Aspects**

- Larger boroughs leave room for new buildings.
- Areas with high population density can be too crowded.
- Very few venues suggest only little entertainment.

### **Assumptions for Cluster Interpretation**

- Buying: affordable price, moderate vibrancy
- Renting: affordable rent, high vibrancy
- Densely populated areas not preferred.
- Very little vibrancy is a drawback.
- Expensive areas not recommended

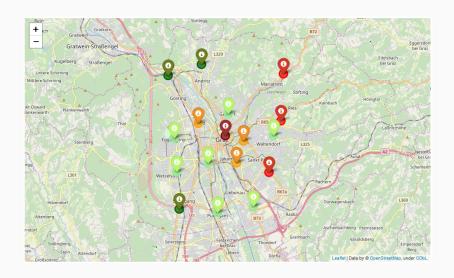
## **Results**

## **Average Cluster Values**

Cluster	Rent EUR/m²	Average Price EUR/m <sup>2</sup>	Area km²	Population Density 1/km²	Venue Count
0	10.94	2566.15	6.11	2941.29	17.57
1	12.90	4642.89	1.16	3287.00	96.00
2	11.66	2363.28	13.68	1097.00	22.00
3	10.53	3184.44	11.00	968.67	8.33
4	10.51	2110.92	3.20	8263.67	12.00

- Largest cluster: cluster 0 (7 boroughs)
- Smallest cluster: cluster 1 (1 borough, inner city)
- Remaining clusters: 3 boroughs each
- Same clusters for rent and price

## Map with Recommendations



## **Discussion**

#### Discussion

#### Results

- Map provides overview and enables further filtering
- Largest cluster: renting
- Dense population: close to transport hubs
- Central borough most expensive
- Reasonable results

#### Method

- Using 4 clusters would combine clusters 2 and 4
- 5 clusters separate periphery and developing boroughs

# Conclusions and Outlook

#### **Conclusions and Outlook**

#### **Conclusions**

- Objective method
- Approach is feasible, provides reasonable results
- Generally applicable to any city

#### Outlook

- More data will lead to better results
- Smaller disctricts to detect hotspots within boroughs
- More available venue data would allow target group-specific analysis