

# Predicting **RESIDENTIAL REAL ESTATE**

In Taipei,Taiwan

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## Workflow

01

**Problem Statement**

02

**Data Acquisition & Preprocessing**

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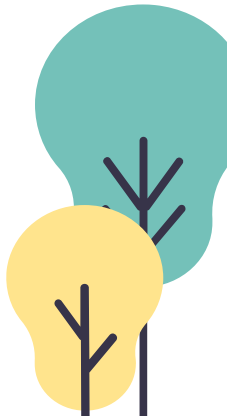
**Predictive Model:  
Linear Model**

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**Socioeconomic Analysis**

06

**Conclusion**



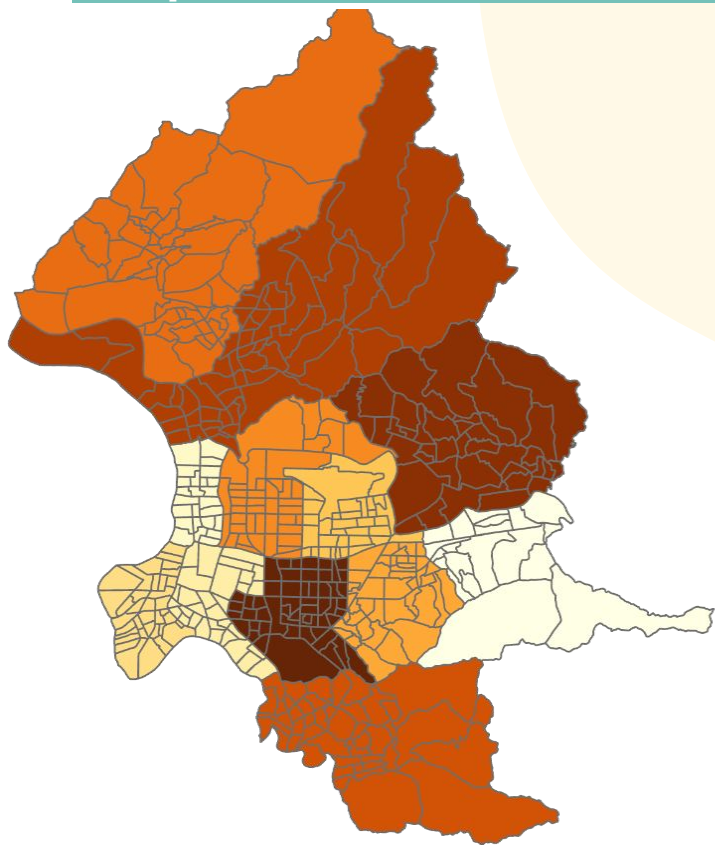
# Problem Statement

- Housing prices around the world have experienced **tremendous increases**, and this has created business opportunities while **worsening inequality**.
- The ability to develop a city that is **balanced in living conditions**, prices, and ability to invest accurately in real estate properties has become essential skills for city governments and its citizens.

Through this project, our team would attempt to evaluate the impact of certain **points of interests (POIs)**, or “**residential attractors**”, on the housing market within the city.



## Taipei in Numbers



### Total Area

271.80 km<sup>2</sup>



### Districts

12



### Density

9,071.42 / km



### Population

2,465,610



### Avg. House Price

US\$ 924,249



## Data Acquisition

**2020**

Housing Sales Price  
Taiwan Ministry of Interior



**40000+** Entries  
**Price & Properties**

**2022**

Point of Interest (POI)  
Google Maps API



**4118** Entries  
**12 POI** Types  
**3 Distances**  
(500m, 1km, 3km)

**2021**

Socioeconomic Factors  
Taiwan Ministry of Interior,  
Ministry of Finance,  
Taipei City Government



**Education, Income** per  
Sub-district,  
**Population** per district

**2021**

Human Activity Flow  
Taiwan Ministry of Interior  
Social Economical GIS (SEGIS)

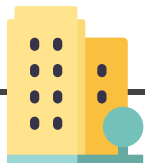


**Phone Signal Data**  
Workday vs. Weekend  
Day vs. night  
(District Level)

# Data Pre-Processing

## Translation

Data CSV file contained  
columns record in  
Traditional Chinese



## Data Cleaning

Certain houses have missing  
information or unique info for  
specific columns



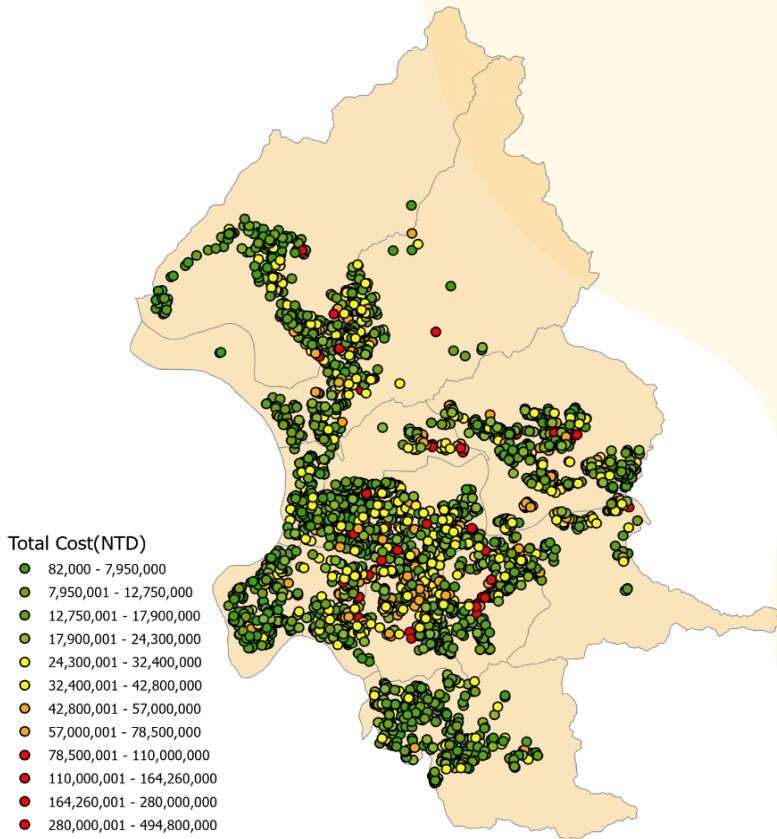
## Data Transformation

Normalization of values,  
One-hot encode



**Ready  
for  
Analysis**

## Data Visualization: Total Price



▣  $\leq$  NTD \$25,000,000  
(USD \$817,750)

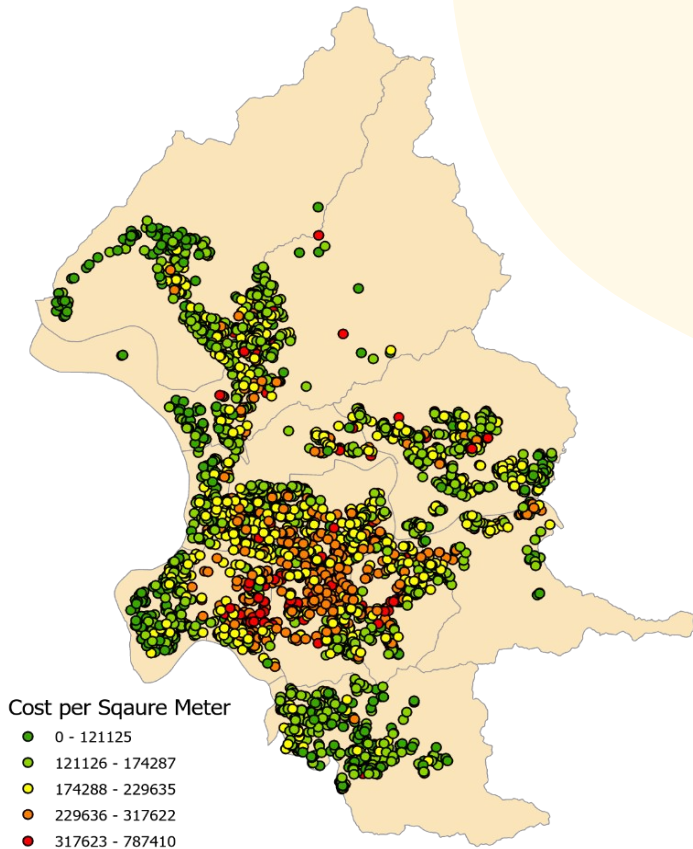
▣  $\geq$  NTD \$80,000,000  
(USD \$2,616,800)

There are a few red spots in most of the districts.

Cluster of medium to expensive houses around city center and sub-centers.

Areas further away from the city center have lower house prices on average.

## Data Visualization: Price per Area



▣  $\leq$  NTD \$200,000

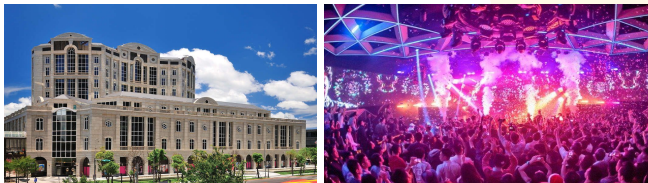
(USD \$6,542)

▣  $\geq$  NTD \$400,000

(USD \$13,084)

This clears up the distribution of prices.  
We can see natural clustering in many districts.

# POIs



Police Station  
Supermarkets  
Hospitals

Bus station  
SubwayStation

University  
Primary School  
Library

Church  
Night Club  
Shopping Mall  
Park



**500 meters**  
**Walking Distance**

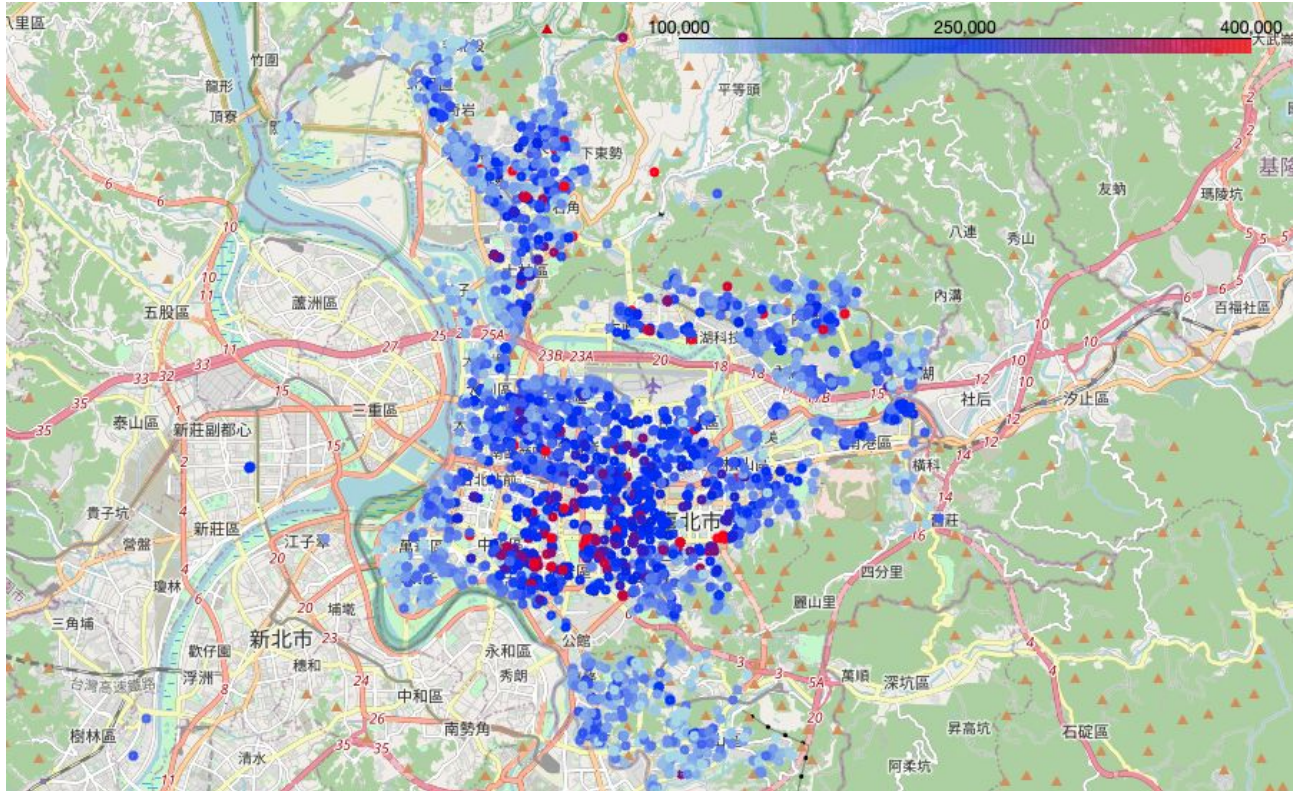


**1 kilometer**  
**Public Transport/  
Scooter Distance**



**3 kilometers**  
**Car Distance**





## Index:

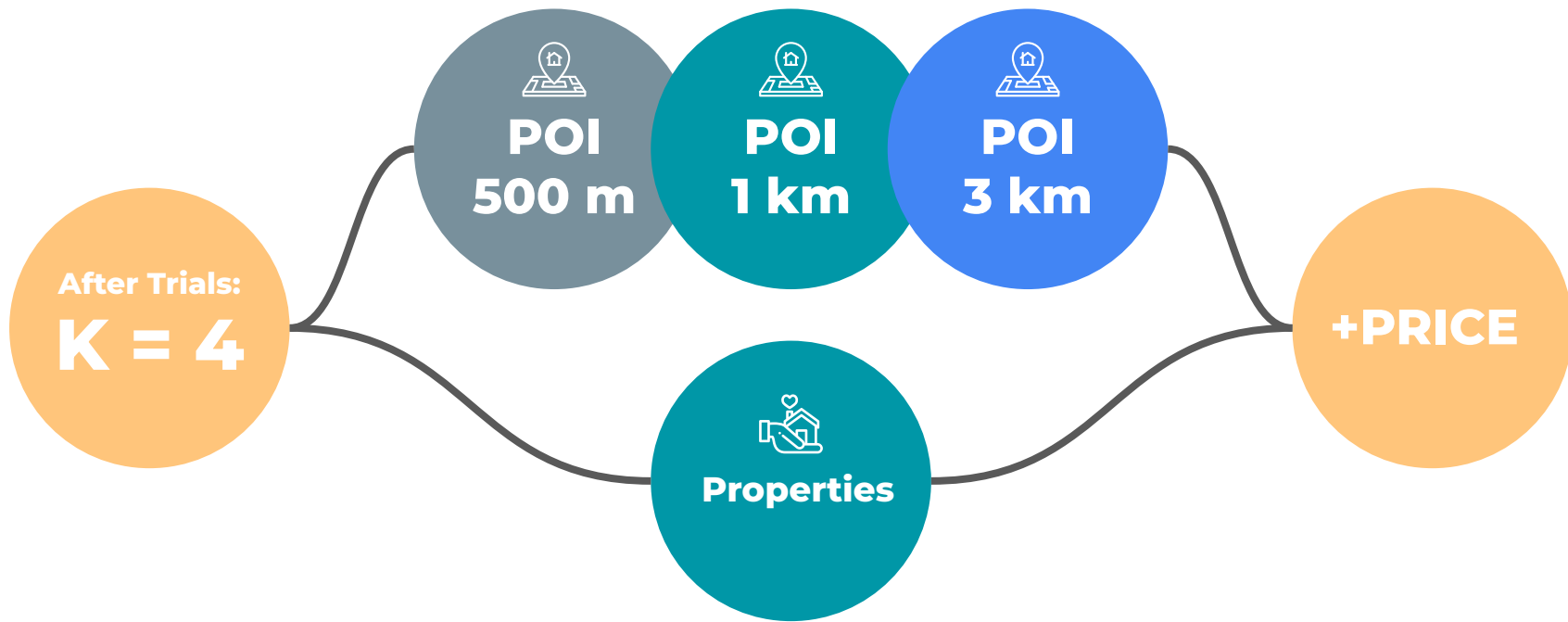
▣ ≤ NTD \$100,000  
(USD \$3,231)

▣  $\geq$  NTD \$400,000  
(USD \$12,925)

This clears up the distribution of prices even more clearly. Colors for each districts and areas are less different.

The red spots have also decreased as this takes the size of houses into consideration.

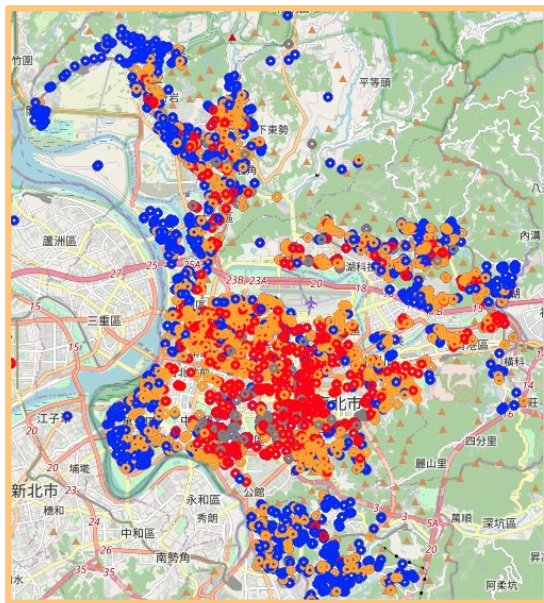
## Clustering: K-Means



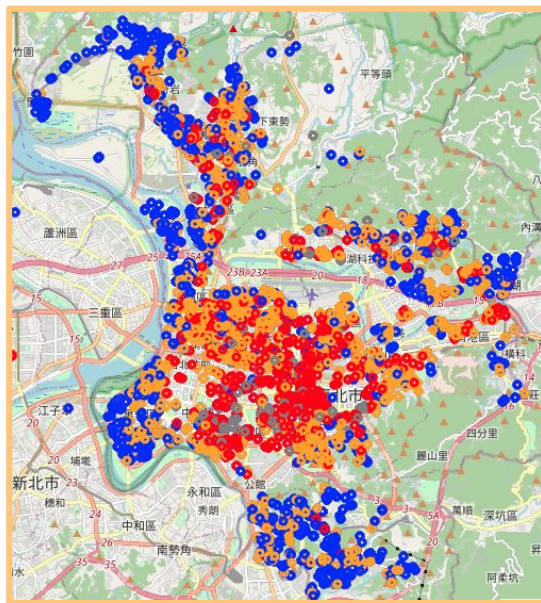
**POI vs  
Properties**

Cluster Legend:  
0: Red, 1: Blue,  
2: Gray, 3: Orange

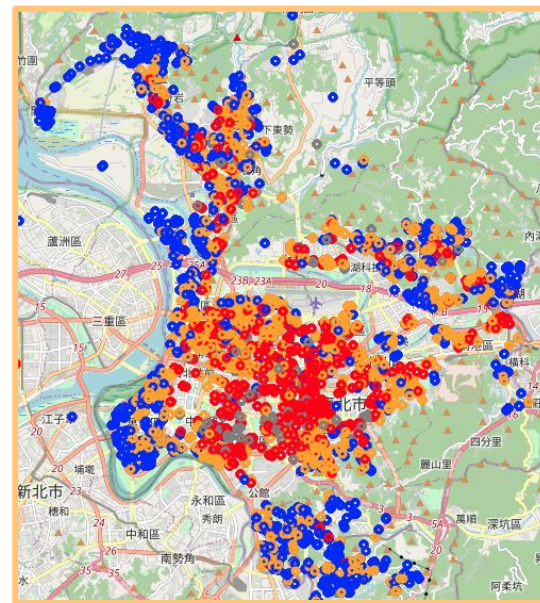
## K-Means: POIs + Price



500m



1 km



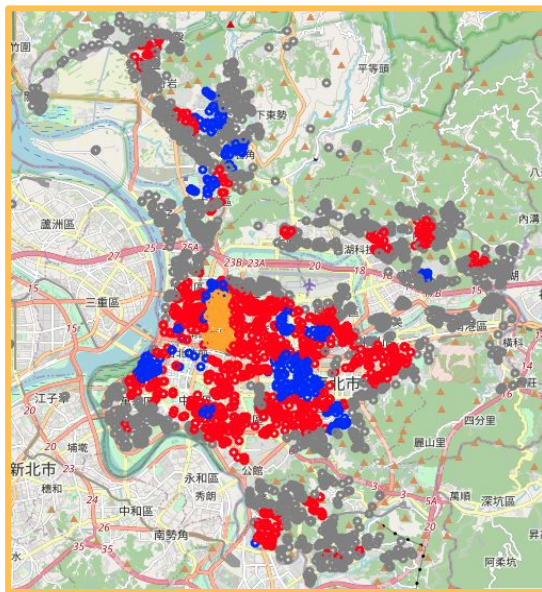
3 km

When price of the house is taken into consideration, it skews the results significantly. Therefore, it should not be included in the clustering model.



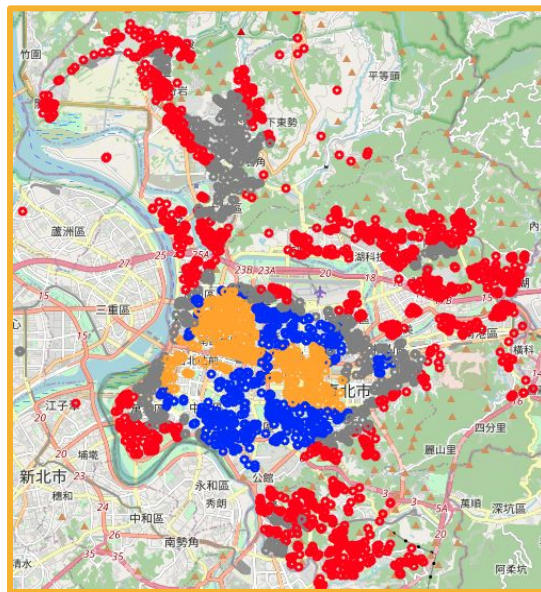
Cluster Legend:  
0: Red, 1: Blue,  
2: Gray, 3: Orange

## K-Means: Only POIs



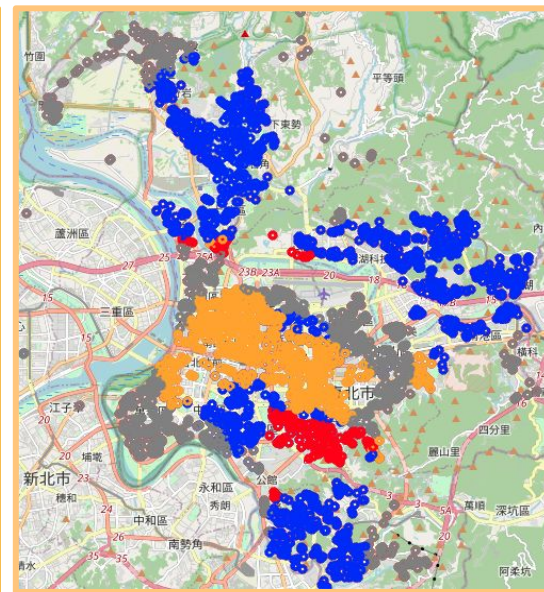
500m

There are small clusters everywhere in the city.



1 km

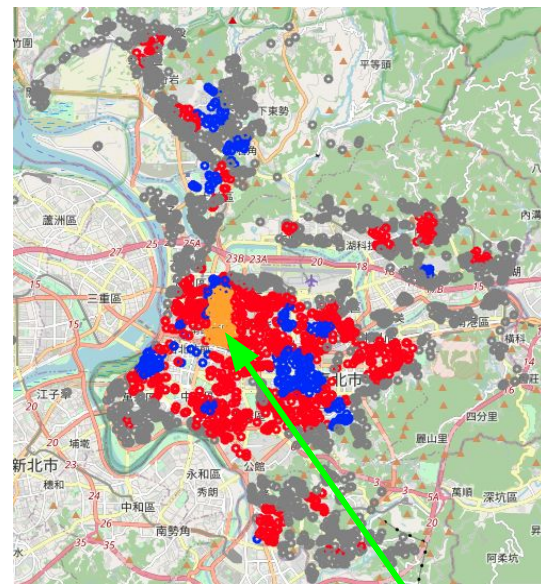
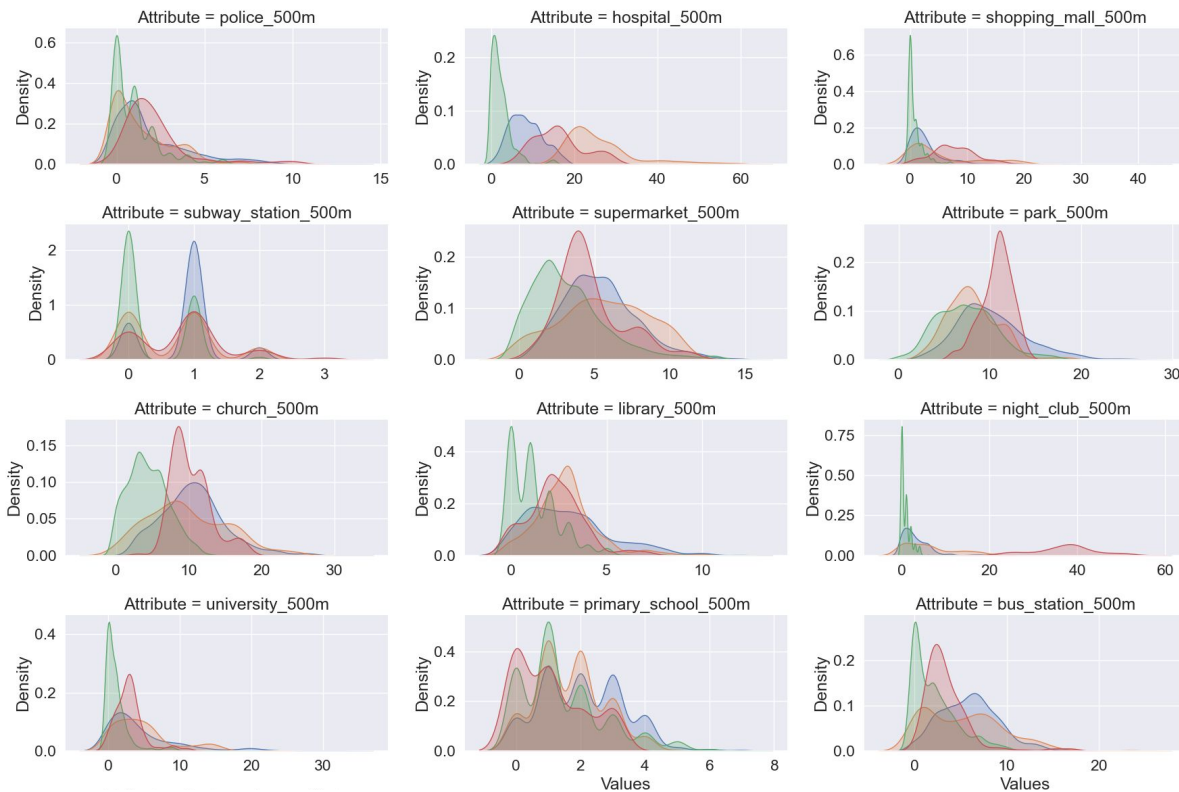
Two main cluster appear in the city, one is old town and CBD.



3 km

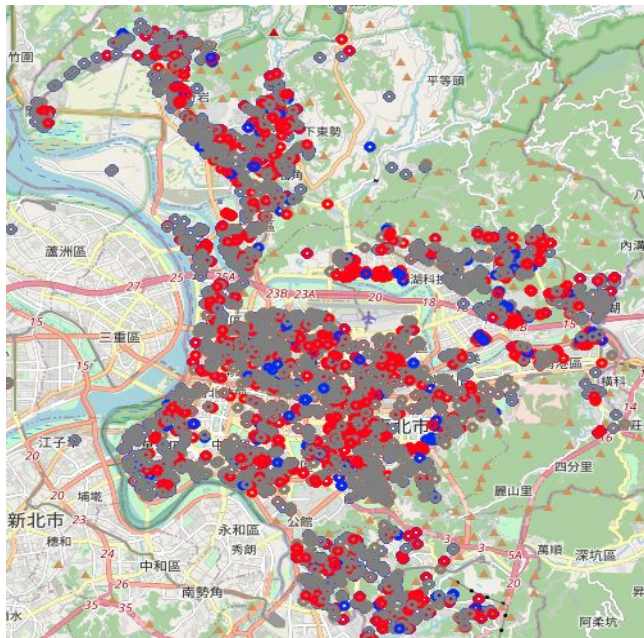
The city is separated into the inner cluster, middle, and outer.

## POI distributions within 500m in different clusters

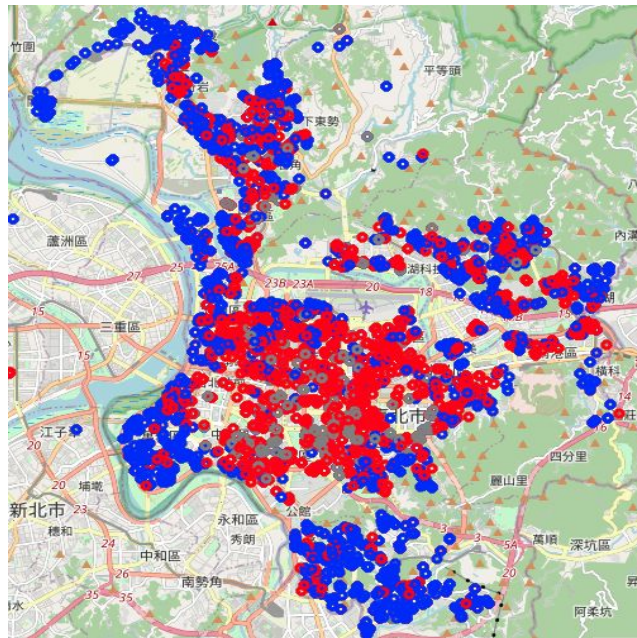




## K-Means: Properties / Properties + Price



**Properties**



**Properties + Price**

### Findings:

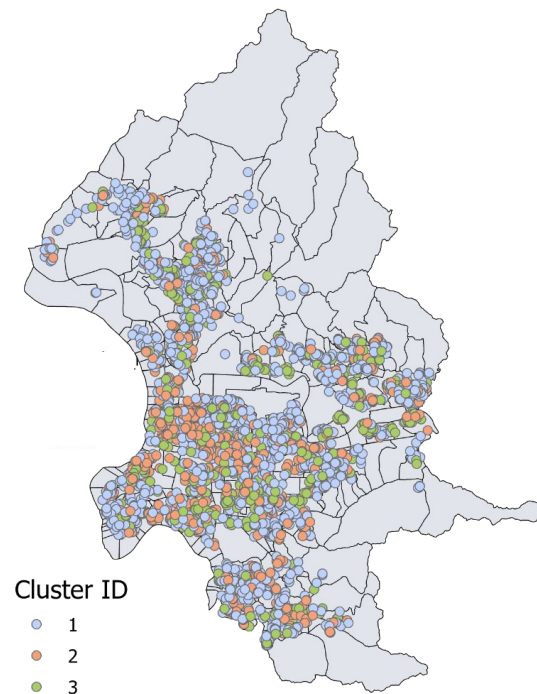
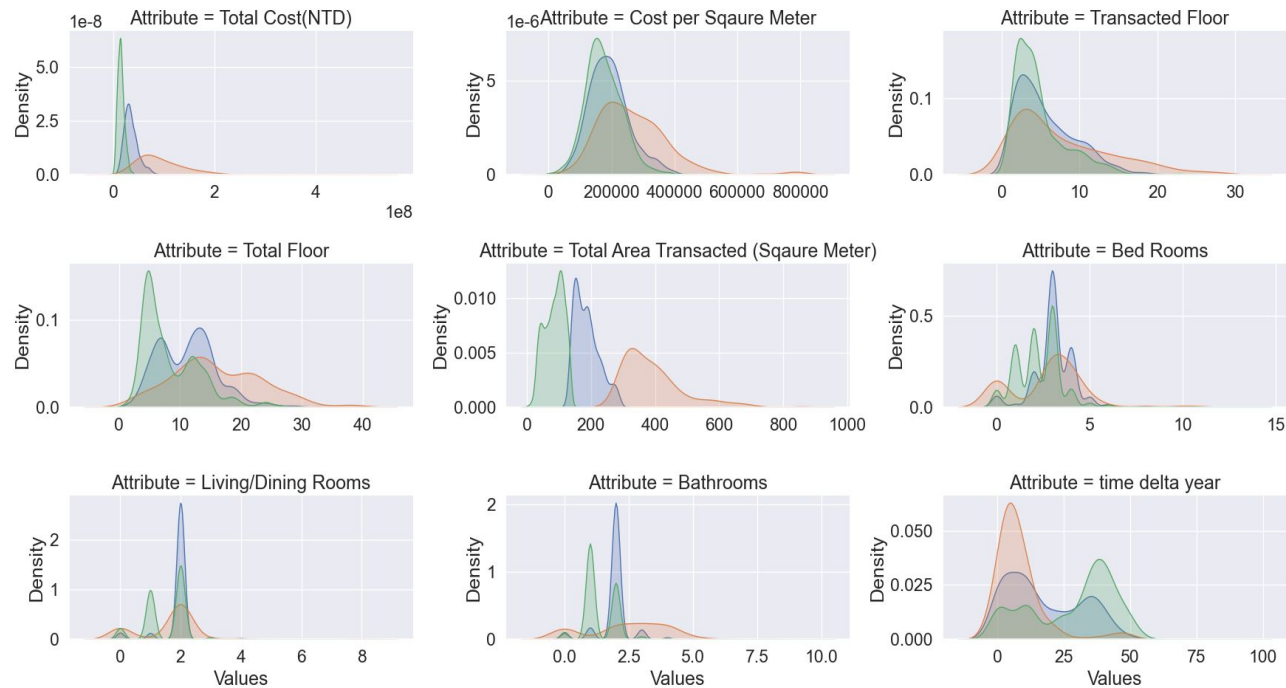
When price is considered, it identifies the city centers and areas outside of city centers.

Therefore, a closer look at the clustering without price is needed.

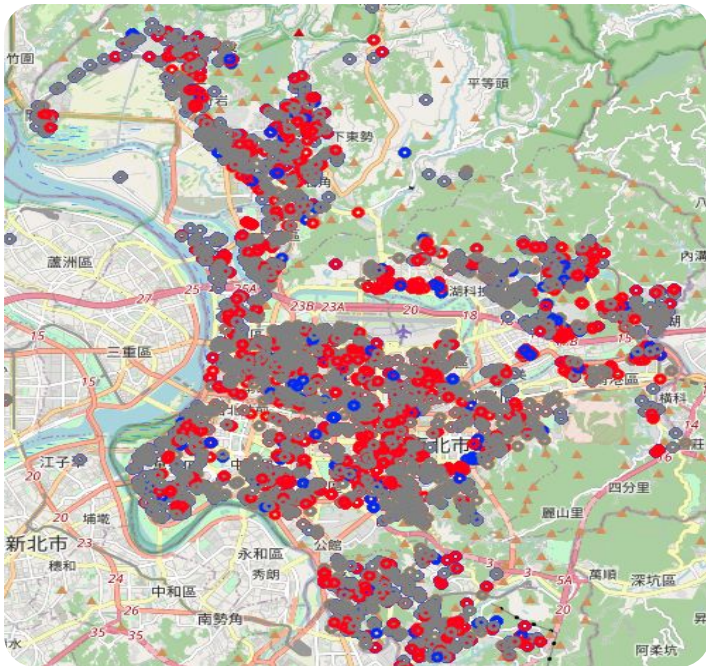
### Cluster Legend:

0: Red, 1: Blue, 2: Gray

## Distributions of properties in each cluster



Cluster Legend:  
0: Red, 1: Blue,  
2: Gray,



## Properties

	Cluster 0	Cluster 1	Cluster 2
Total Cost	34458925	99205981	14354527
Cost m2	193140	269623	175929
Floor	5.6845	7.74	4.80
Total floor	11.318	16.051	8.4
Total Area	186.191	391.259846	83.94
Bed rooms	3.079381	2.77	2.24
Liv/Dine Rm	1.9020	1.64	1.49
Bathrooms	1.948454	2.46	1.39
Age	17	8.32	28

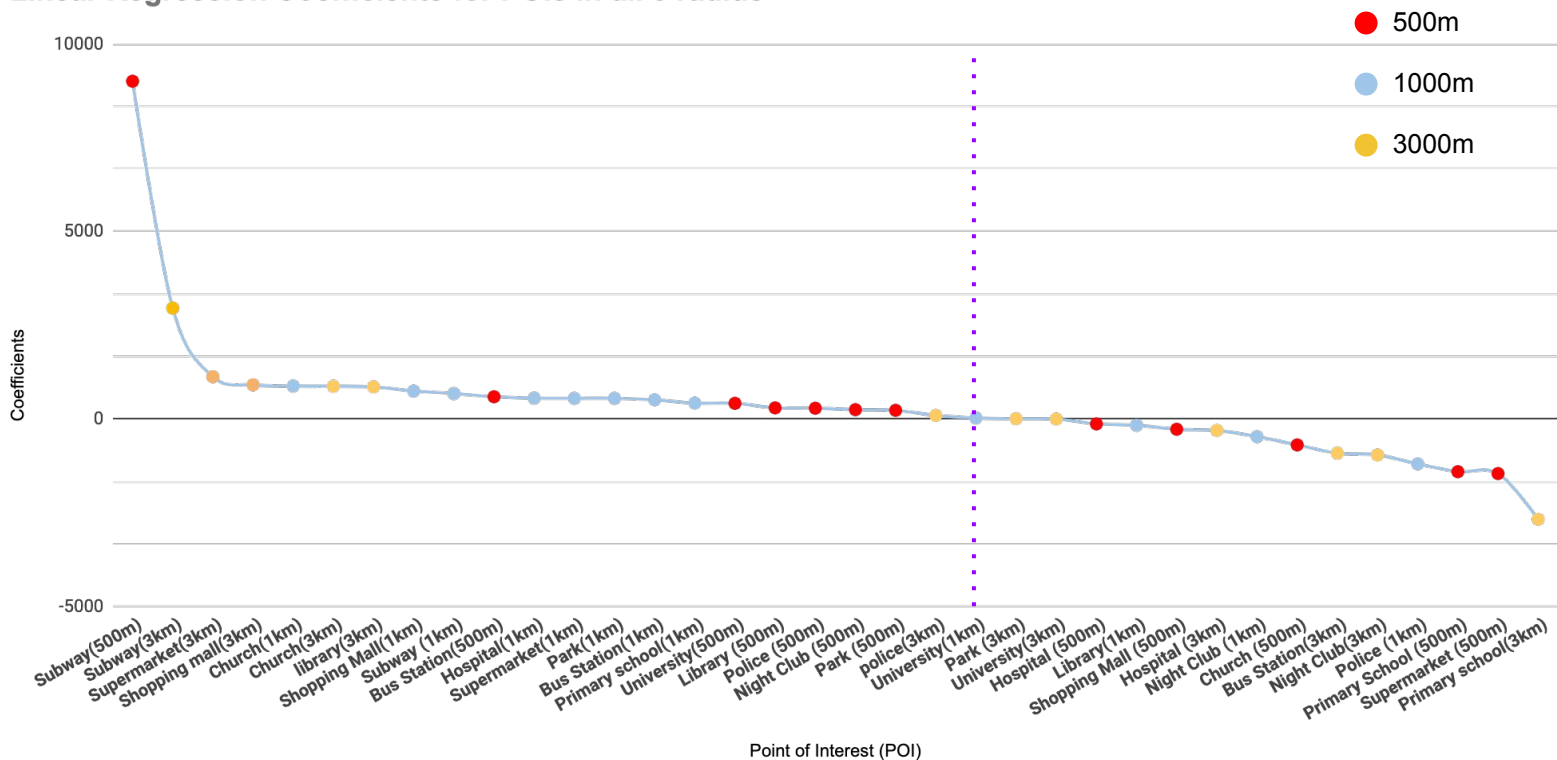


## Linear Regression Model

$$\textit{Unit Price} = \sum_i \alpha_i * F_i + \textit{bias}$$

# Linear Regression Model — All POI in 3 Radius

Linear Regression Coefficients for POIs in all 3 radius



## Linear Regression Model — POI

### Findings:

- **Subway stations** and **Bus stations** are important in shorter distances, and becomes less important further away.
- At 3 km, **library, church, and shopping malls** has the highest importance since these POIs may only be occasional visits.
- **Park** and **University**'s importance would decrease as it becomes further away from the house.
- **Hospitals** generally have a limited positive influence.
- **Nightclubs** influence is still unclear, as the influence does not correlate with the distance.
- **Supermarkets, police, and primary schools** have negative influence, since these POIs may be prevalent in most density areas.

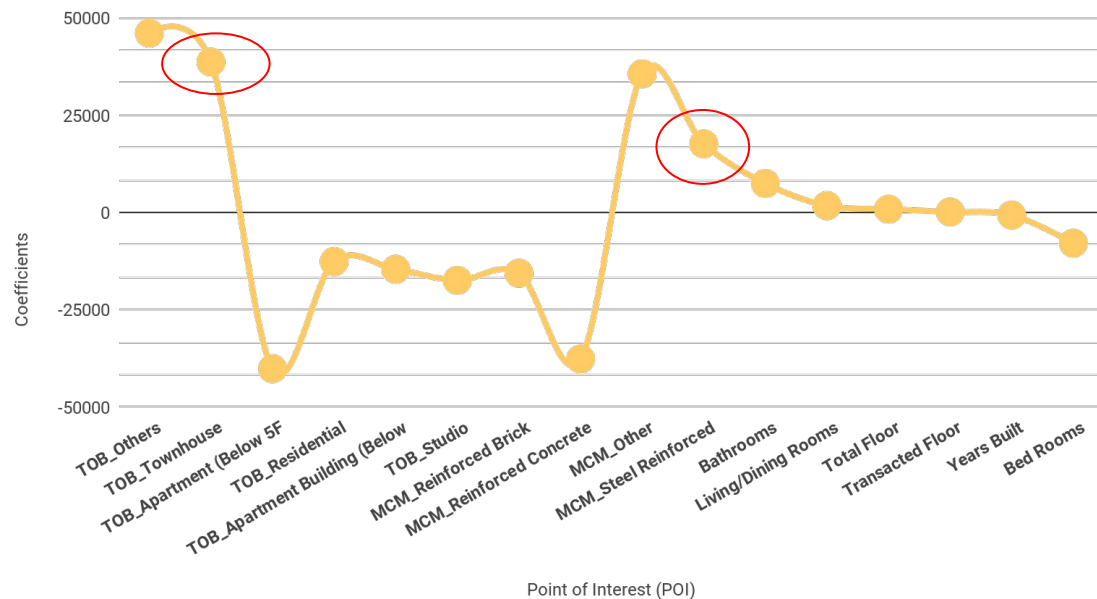
Linear Regression Coefficients for POI in 3 Radius





# Linear Regression Model — House Properties

Linear Regression Coefficients for House Properties

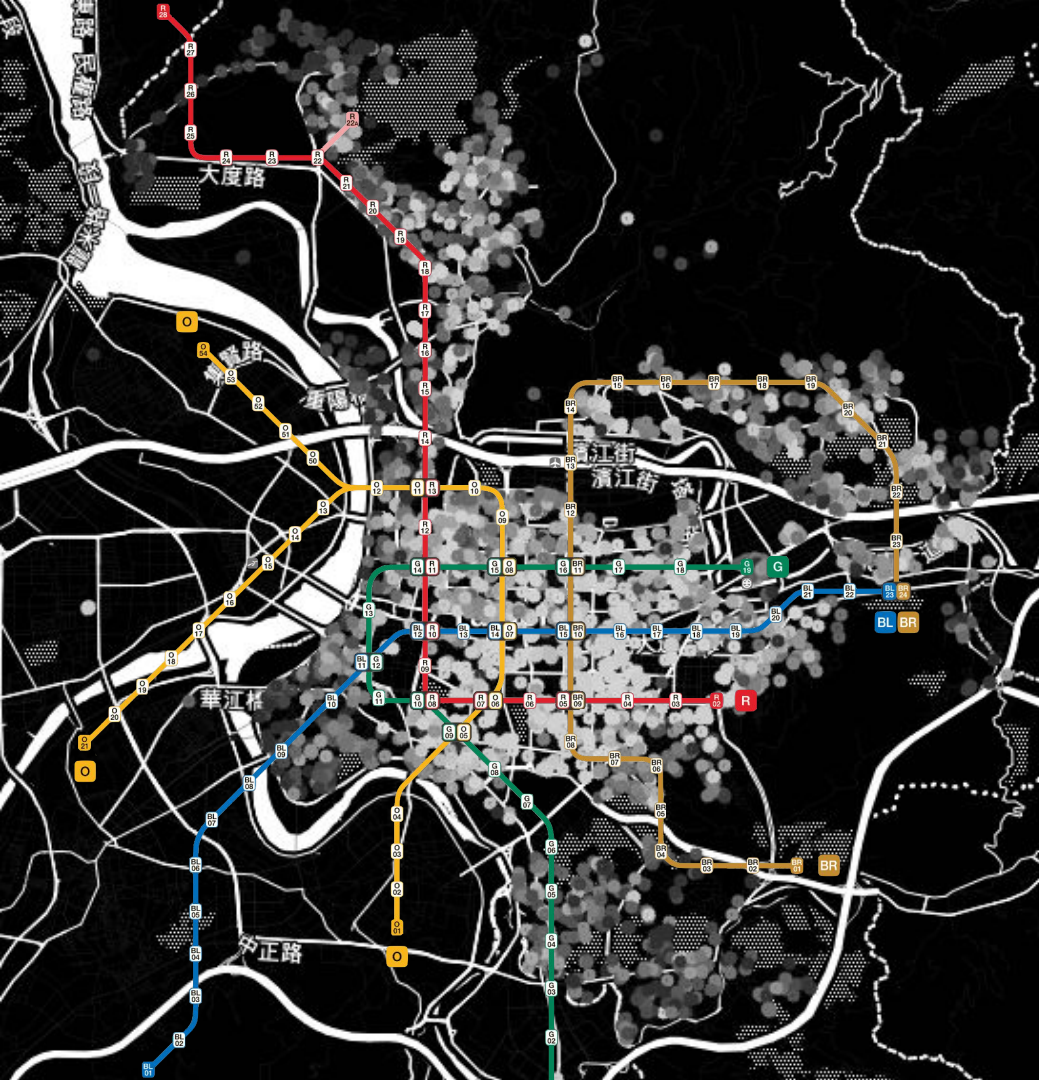


TOB → Type of Buildings

MCM → Main Construction Method

## Findings:

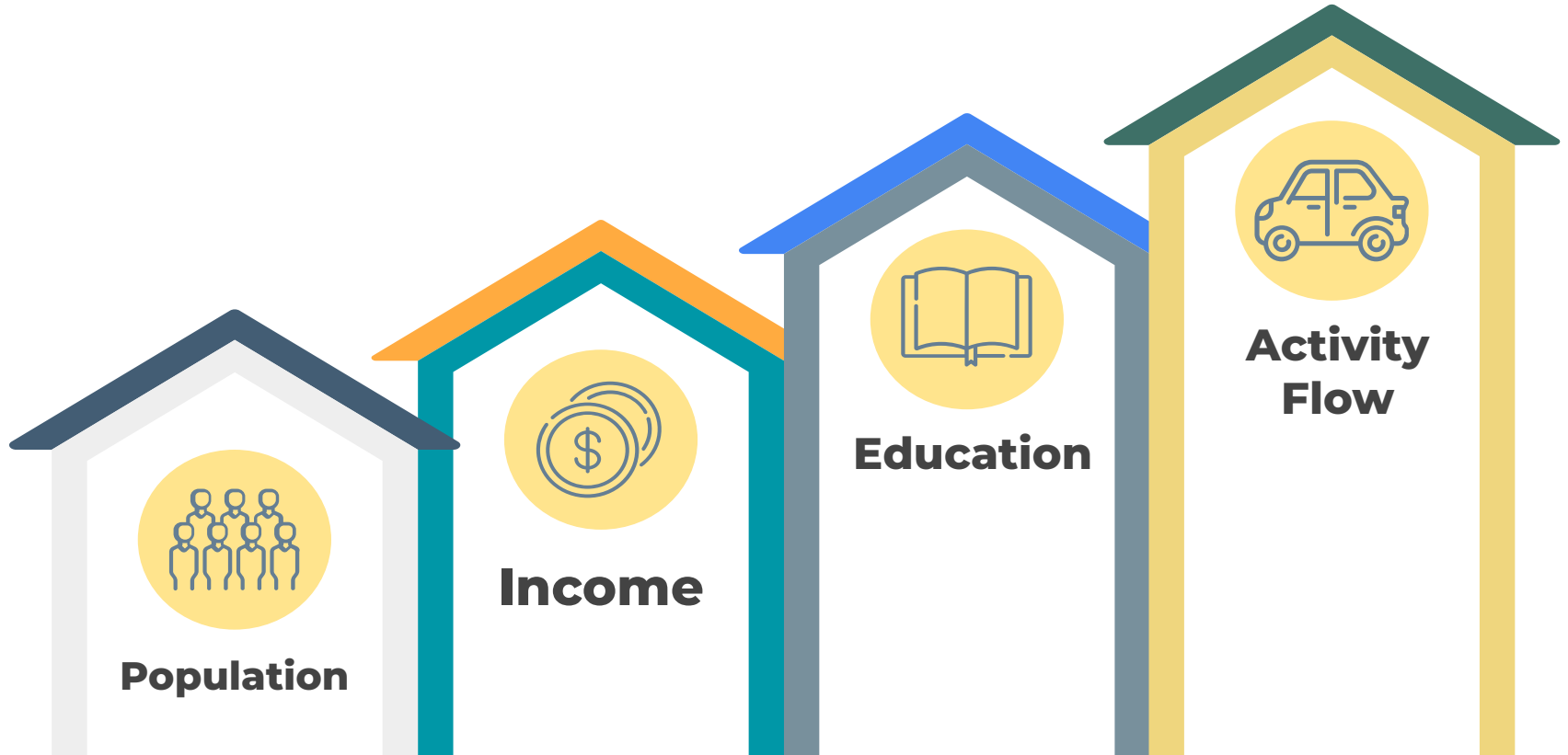
- **TOB\_Others** include office buildings, factories, and storefront that are registered as residential, so it could be considered as an outlier.
- **Townhouses** are rare in Asian cities and higher priced as seen in the model. Other types of buildings does not have large influence as townhouse does.
- **Steel Reinforced Concrete(SRC)** has a large influence compared to other factors, and as most houses built are **Reinforced Concrete(RC)** it has less influence.



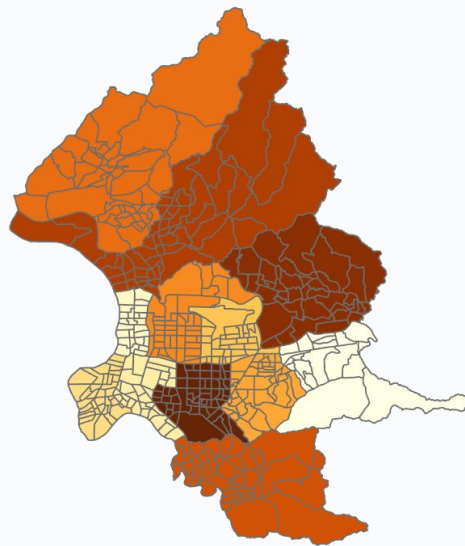
**In- Situ  
Validation:**  
Subway Station's  
High importance  
seen on real map

Lighter areas on this map is most expensive.

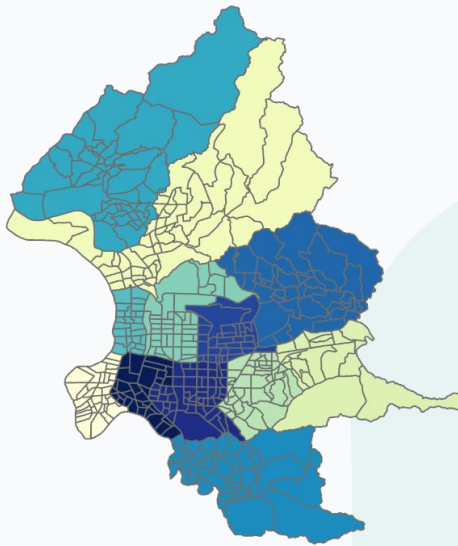
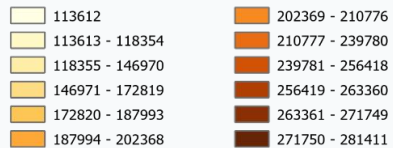
## Socioeconomic Factors



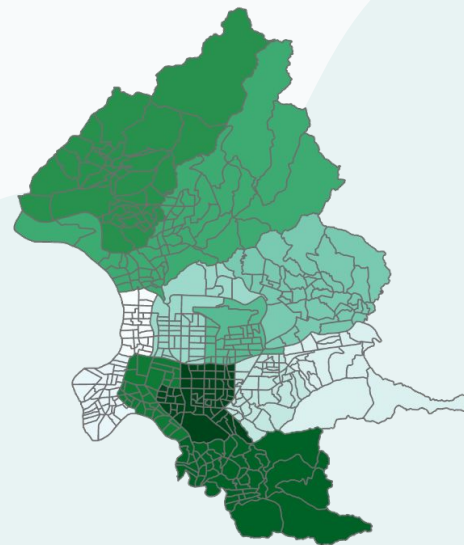
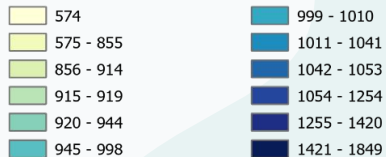
## Socioeconomic Factors Overview



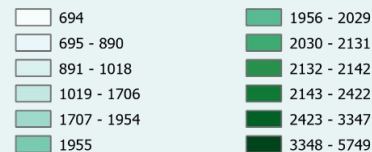
**Population**



**Avg Income**



**Doctorate degree**



# Average Income vs. Total House Price

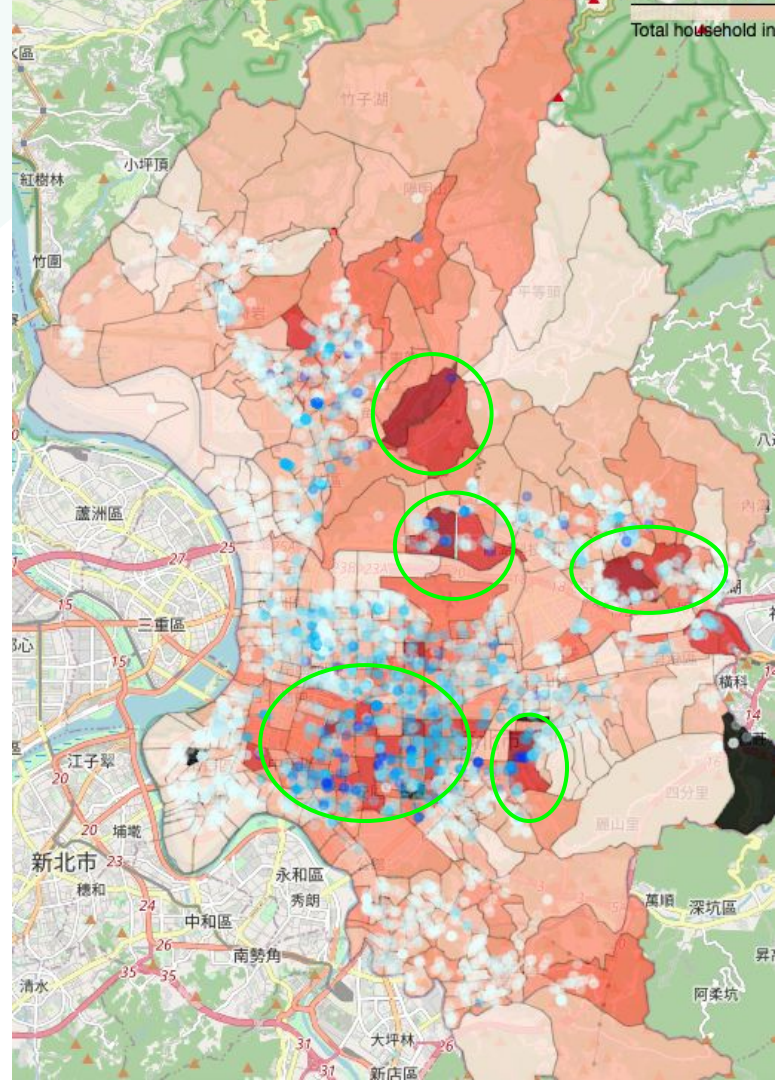
Houses with higher cost per square meter tends to land in places with higher average income.

Average Income in NTD in Thousands

516 791 1,066 1,341 1,615 1,890 2,165

Cost per Square Meter

100,000 160,000 220,000 280,000 340,000 400,000



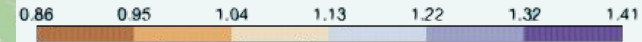
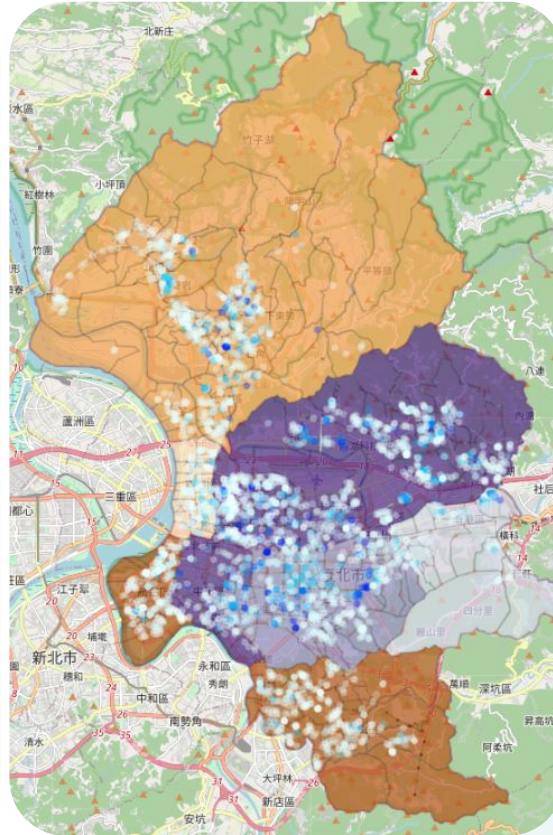
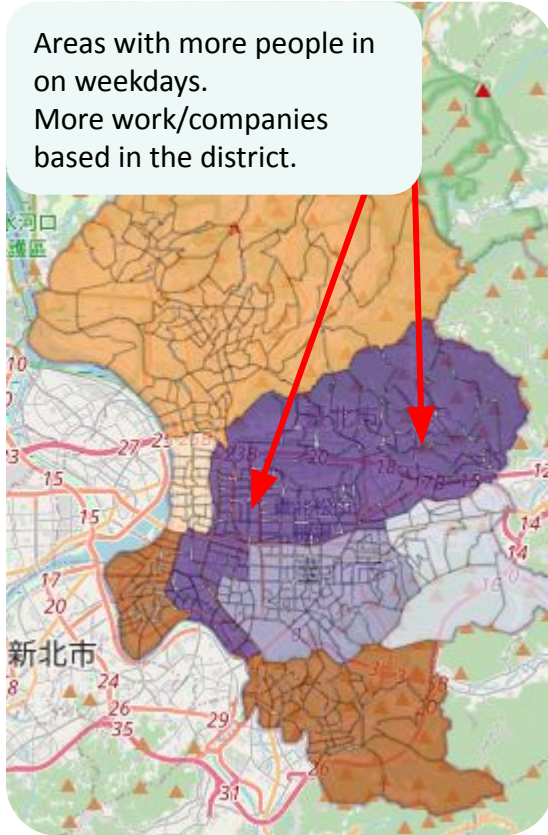


## House Price v.s. Income Correlation



## Activity Flow Day Time

Areas with more people in on weekdays.  
More work/companies based in the district.



### Weekday Daytime Count Weekend Daytime Count

If ratio:

- > 1: Higher Weekday Count
- = 1: Small or No Differences
- < 1: Lower Weekday Count

Therefore:

- >1 would represent more people working, and <1 would be more people living there or leisure zones.

### 1. Transit-oriented Development (TOD):

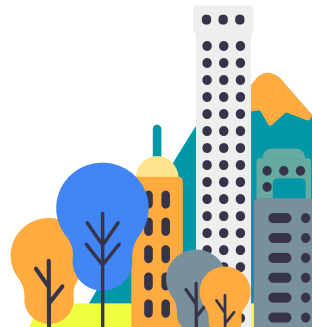
- Transportation POIs, subway stations, and bus stations has high importance
  - TOD development methodology help create a more sustainable city.
  - **Walk more to save your wallet!**
- POIs for leisure (shopping mall) are important if there are at a certain distance (3km).
- Good city structure makes common POIs (hospital, park, universities) less important.

### 2. House Properties:

- Important but not as POIs do.

### 3. Socioeconomics:

- **Earn more, liver better**
- Working areas have the most expensive houses
  - **Spend some time commuting to save money!**





● ● Data is not free!

November 4, 2022

● CIVENG 263N **\$22,483.03**

● Bob API **\$0.00**

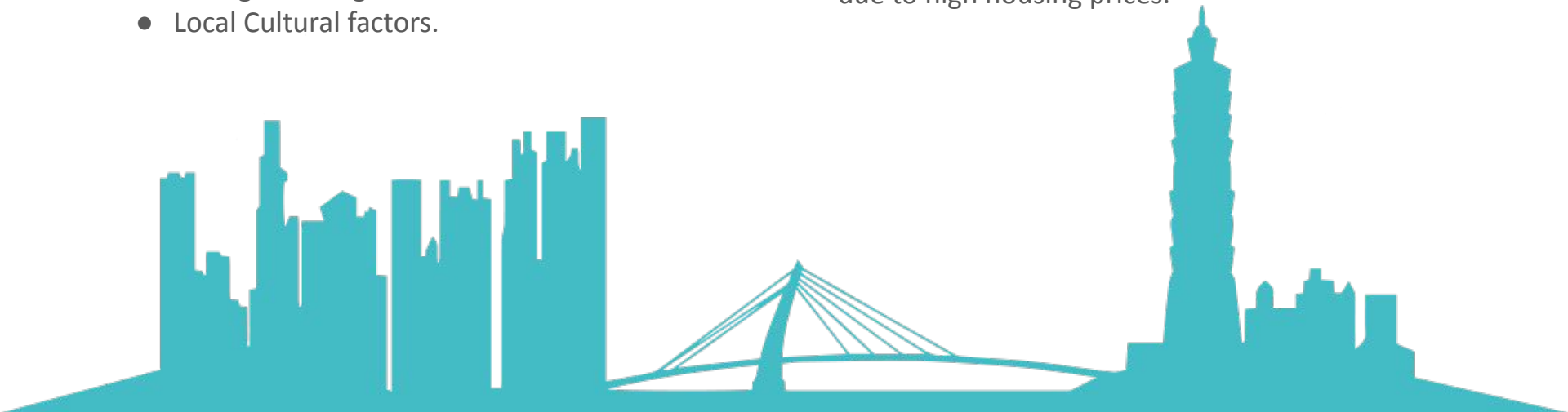
## Challenges, Caveats, and Opportunities

### Challenges & Caveats

- Cost of Google Maps POI Query
  - USD \$22,000 Bill
- POIs change over time.
- Data are not from the same year.
- Certain data have been dropped due to large missing entries.
- Local Cultural factors.

### Opportunities

- Sustainable city development.
- Improve equality, housing access and purchase power.
- Reduce further urban sprawl due to high housing prices.



**Thank you!**

Questions?



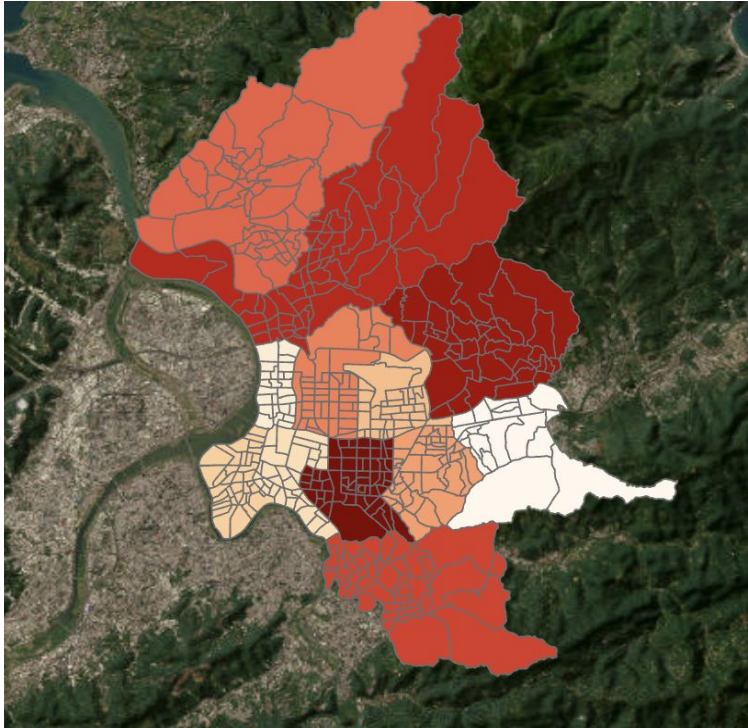


# Appendix Slides

Questions?



## Taipei City Background Information



<b>Administrative Districts:</b>	12
<b>Total Area:</b>	271.80 km <sup>2</sup>
Compared to San Francisco	121.0 km <sup>2</sup>
<b>Household Registered Population</b>	2,465,610
<b>Metropolitan Area Population:</b>	6,894,643
<b>Population Density:</b>	9,071.42 / km
<b>GDP Per capita</b>	\$35,513
<b>Purchasing power parities (PPP)</b>	US\$ 760,00 in 2019
<b>Average House Price</b>	US\$ 924,249 in 2020

# Point of Interest List & Background Information



POI	Total Number	Remarks
Police Station	104	
Hospital	36 Hospital 3,541 Clinics	478 in Taiwan
Supermarket	211	
Bus Station	280 Lines, 2437 stops	
Subway Station	131	6 Lines
Library	76	
University	28	
Primary School	151	142 Public, 9 Private
Church	420	Buddhist/Tao Temple: 1857
Nightclub	N/A	Definition Uncom
Shopping Mall	62	1 per 88k population
Park	616	



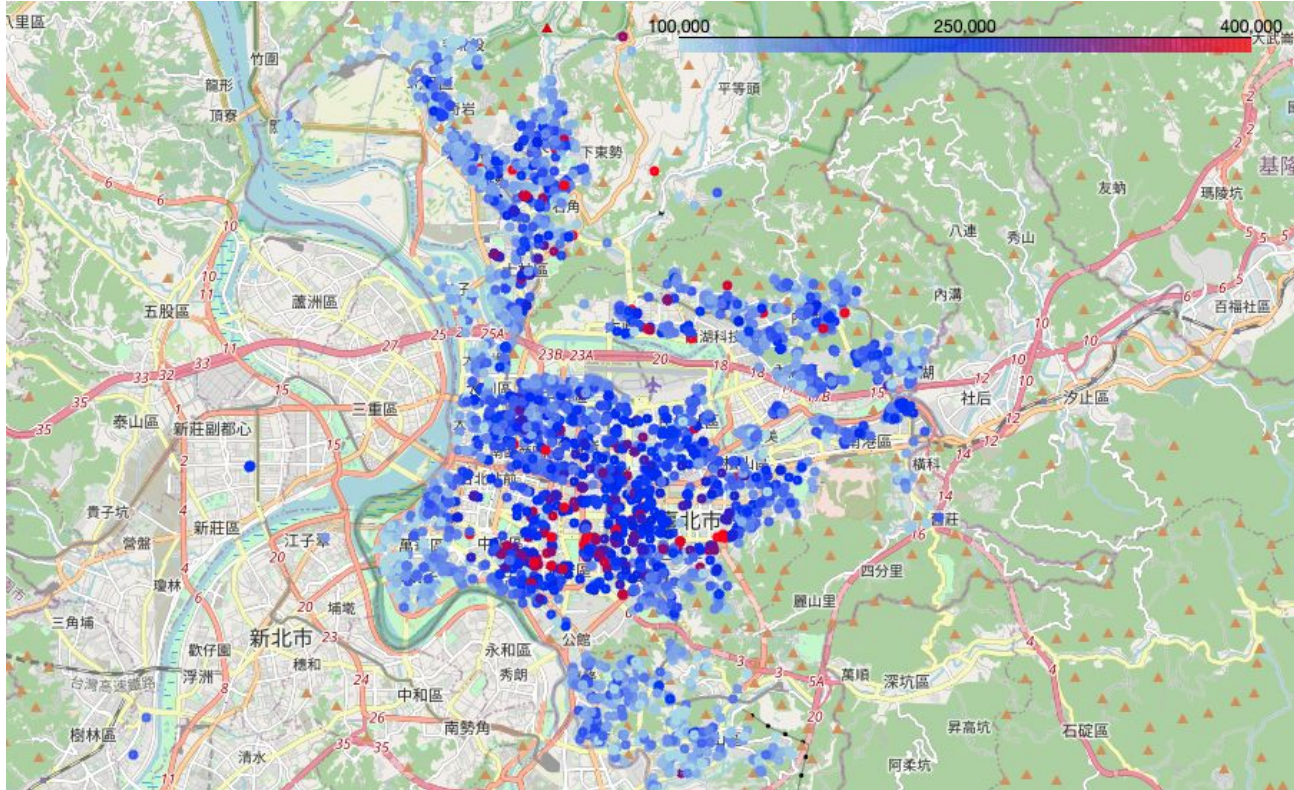
▣ ≤ NTD \$5,000,000  
(USD \$161,574)

■ ≥ NTD \$80,000,000  
(USD \$2,585,184)

There are a few red spots in most of the districts.

Cluster of medium to expensive houses around city center and sub-centers.

Areas further away from the city center have lower house prices on average.



## Index:

- ▣ ≤ NTD \$100,000  
(USD \$3,231)

▣ ≥ NTD \$400,000  
(USD \$12,925)

This clears up the distribution of prices even more clearly. Colors for each districts and areas are less different.

The red spots have also decreased as this takes the size of houses into consideration.



# Average Income vs. Total House Price

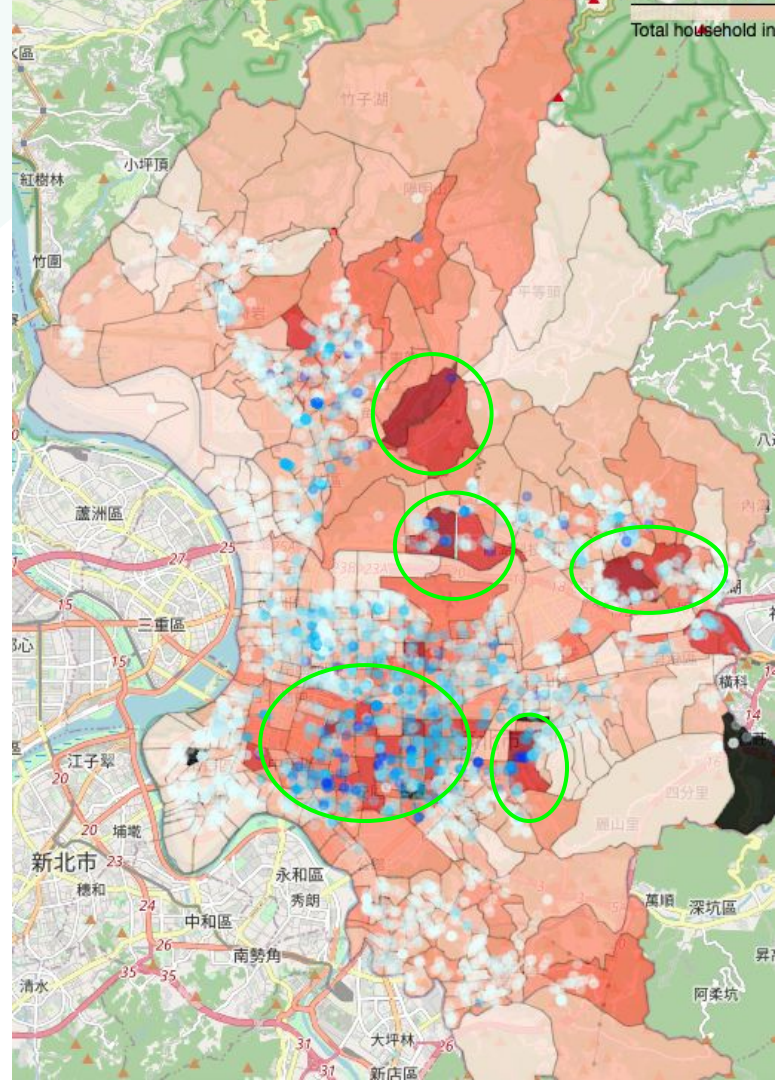
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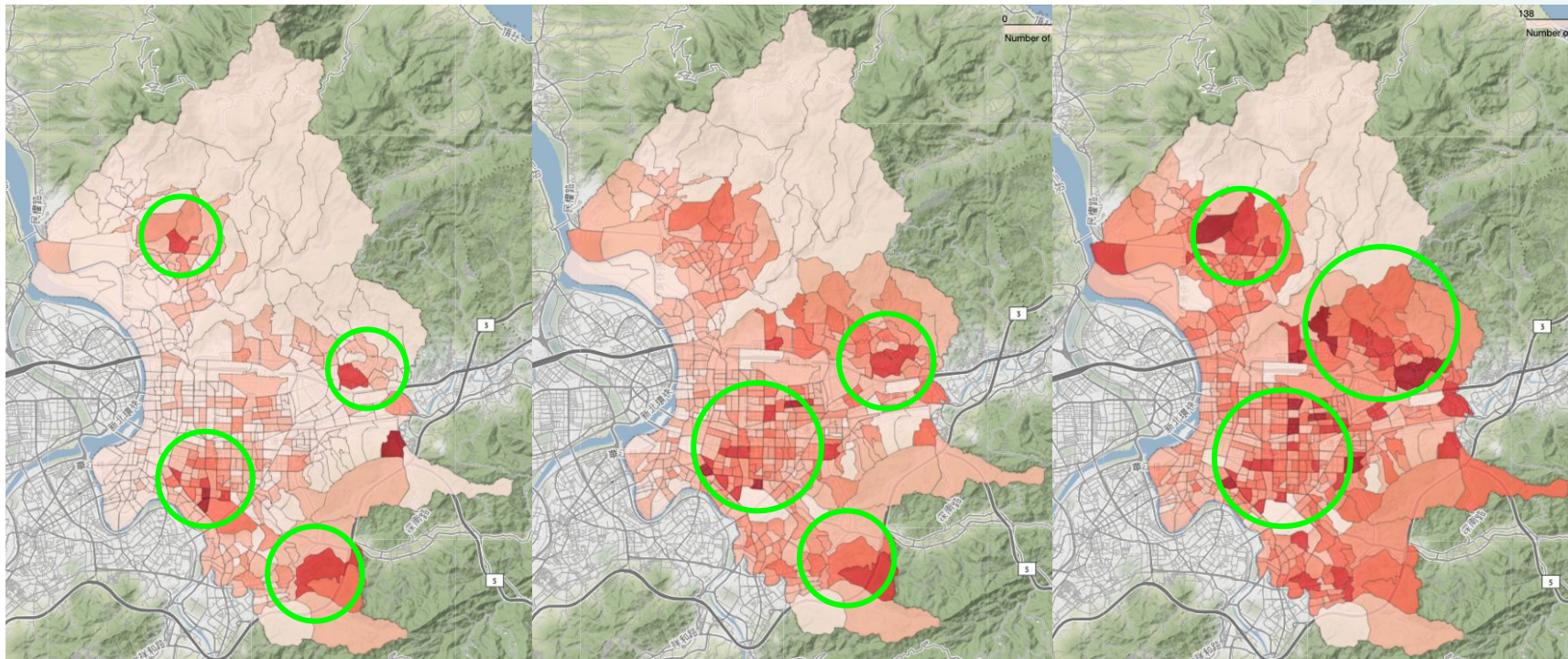
100,000 160,000 220,000 280,000 340,000 400,000



## House Price v.s. Income Correlation



## Education



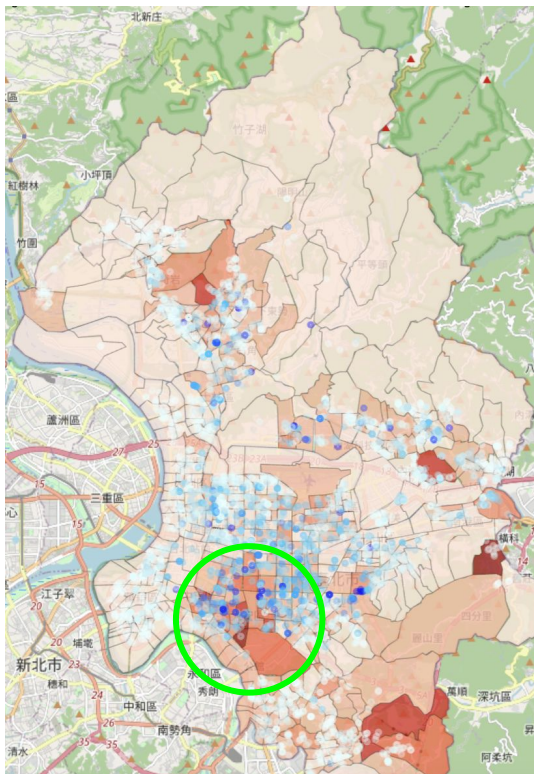
- **Doctorate**

- **Masters**

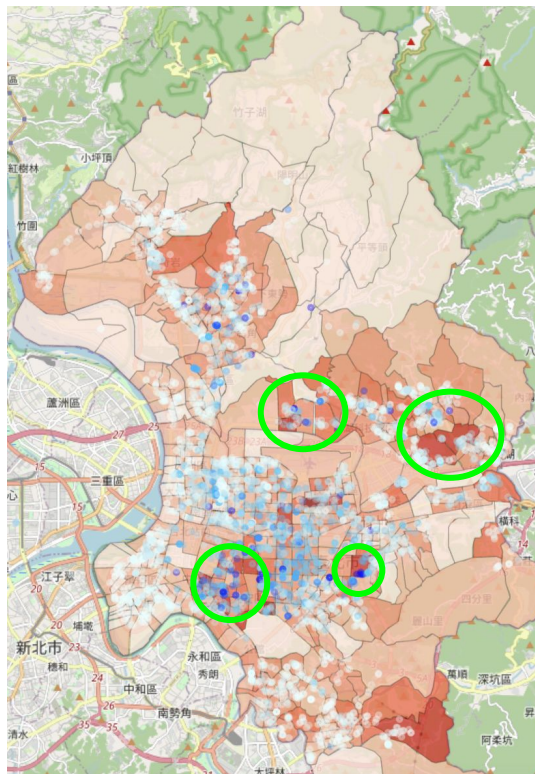
- **Undergrad**



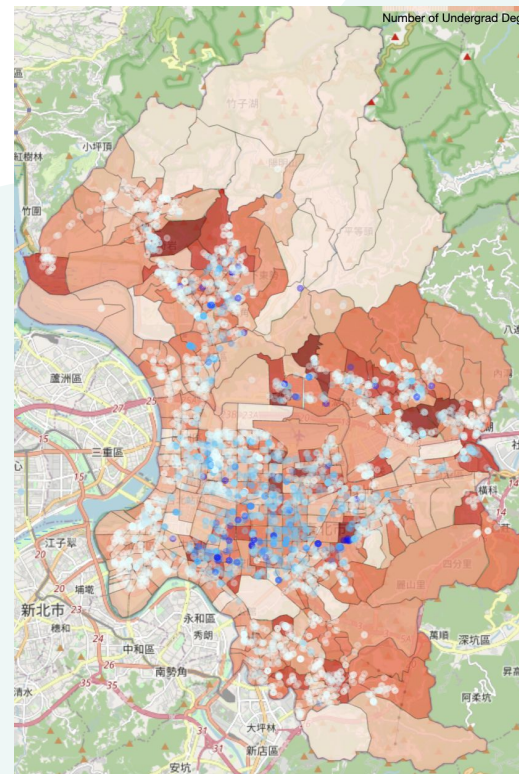
## Education v.s. House Price



• **Doctorate**

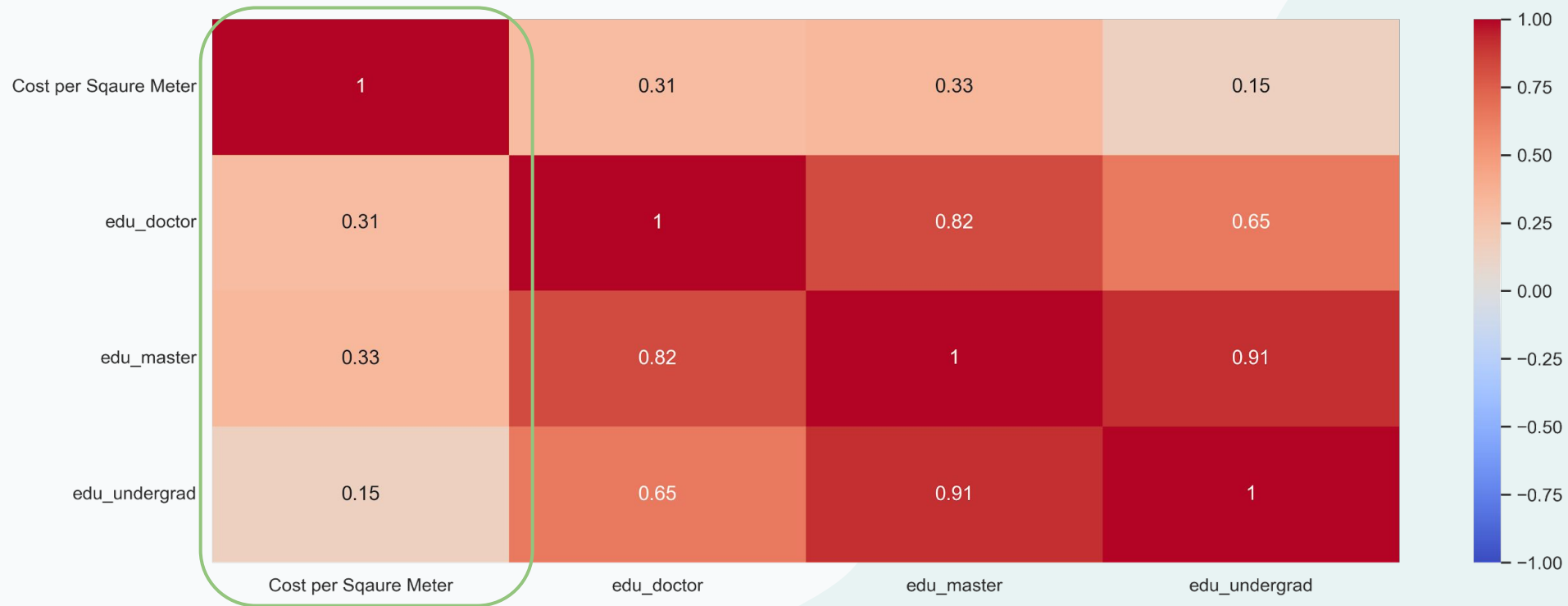


• **Masters**



• **Undergrad**

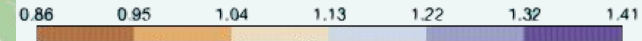
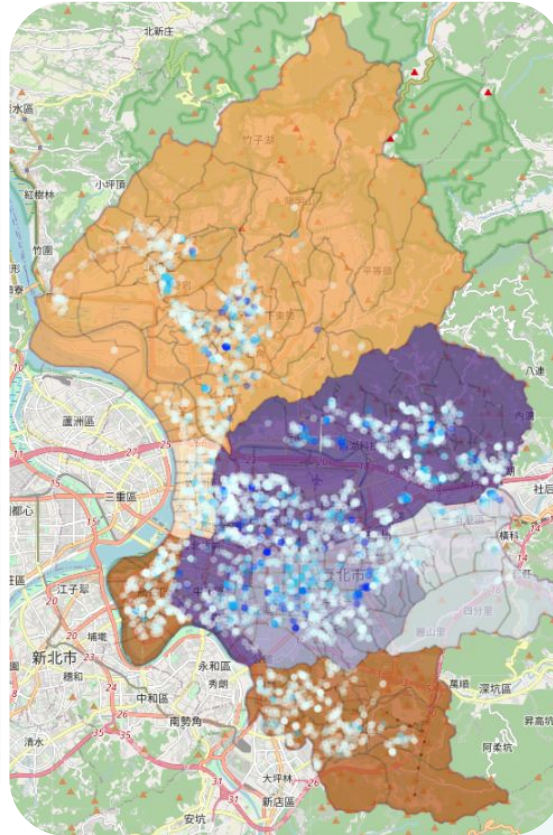
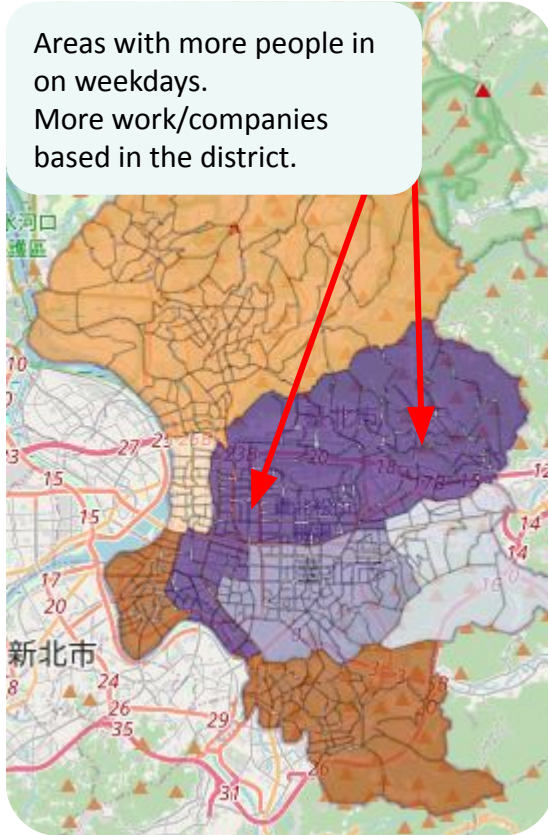
## House Price v.s. Education Correlation





## Activity Flow Day Time

Areas with more people in on weekdays.  
More work/companies based in the district.



### Weekday Daytime Count Weekend Daytime Count

If ratio:

- $> 1$ : Higher Weekday Count
- $= 1$ : Small or No Differences
- $< 1$ : Lower Weekday Count

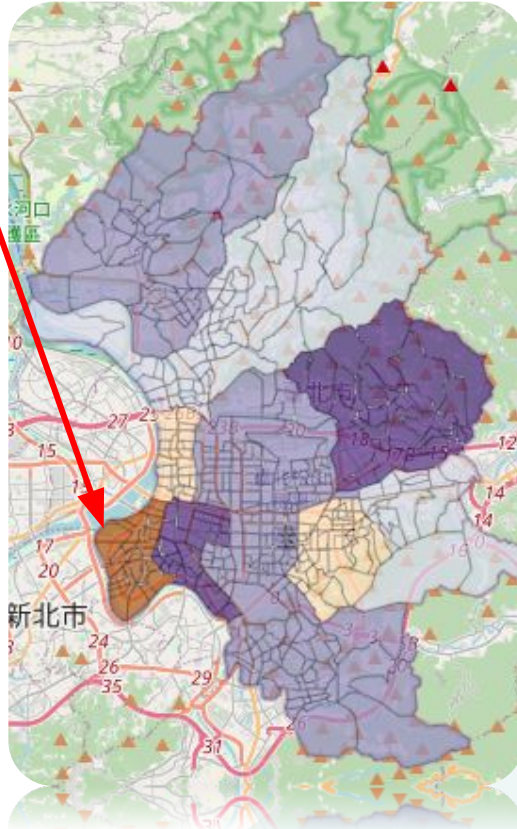
Therefore:

- $> 1$  would represent more people working, and  $< 1$  would be more people living there or leisure zones.

## Activity Flow Night Time

Differences are smaller, but areas with more weekend visits are identified.

Overall, the night time activity count has a smaller difference compared to day time.



0.987 0.994 1.001 1.008 1.016 1.023 1.030

### Weekday Night time Count Weekend Night time Count

If ratio:

- $> 1$ : Higher Weekday Count
- $= 1$ : Small or No Differences
- $< 1$ : Lower Weekday Count

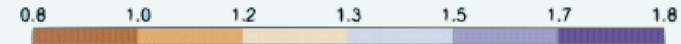
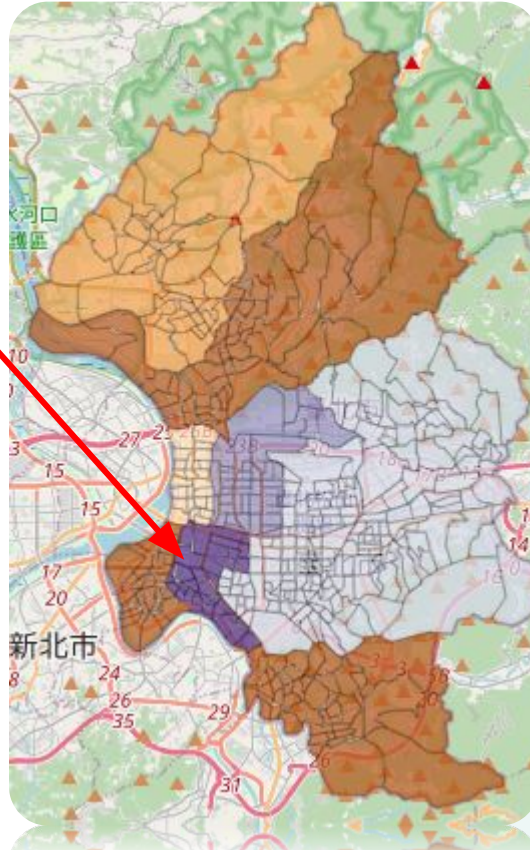
Therefore:

- $> 1$  would represent more people working, and  $< 1$  would be more people living there or leisure zones.

## Activity Flow Workday Day vs. Night

Overall, this shows the the city faces a large inflow of workers in the day time.

The more purple the district is would indicate more people work there and less people live there.



### Weekday Day time Count Weekday Night time Count

If ratio:

- $> 1$ : Higher Weekday Count
- $= 1$ : Small or No Differences
- $< 1$ : Lower Weekday Count

Therefore:

- $> 1$  would represent more people working, and  $< 1$  would be more people living there or leisure zones.