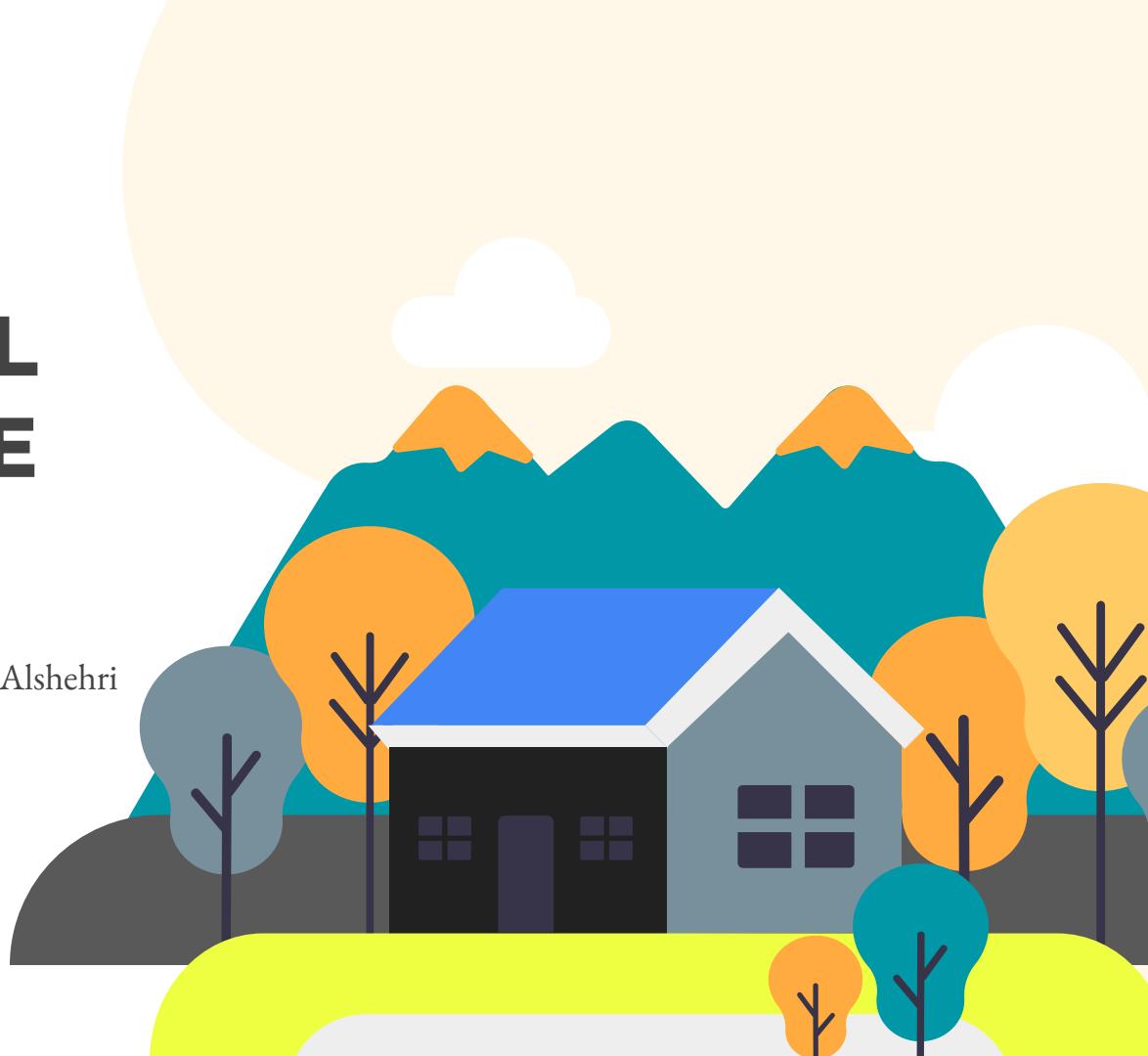


Predicting **RESIDENTIAL** **REAL ESTATE**

In Taipei, Taiwan

Morris Chang, Eric Yuxin Miao, Ibtihal Alshehri





Workflow

01

Problem Statement

02

Data Acquisition & Preprocessing

03

Cluster

04

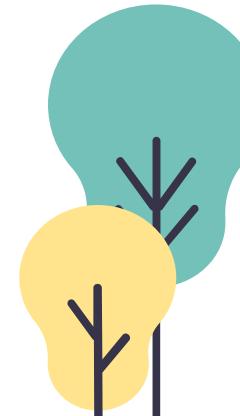
Predictive Model:
Linear Model

05

Socioeconomic Analysis

06

Conclusion



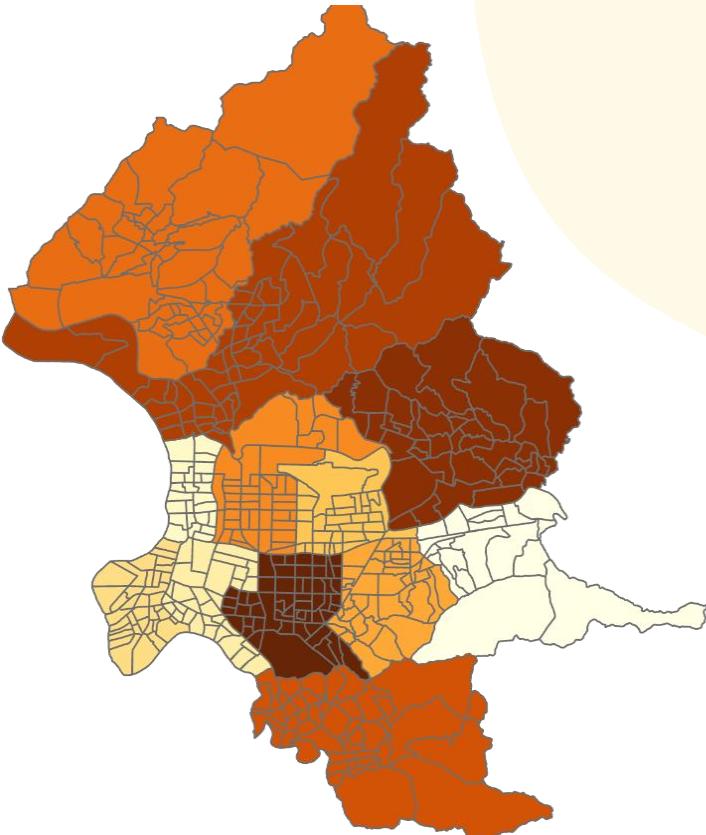
Problem Statement

- 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 km²



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



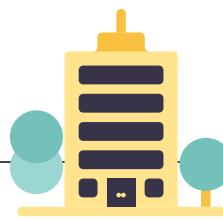
40000+ Entries
Price & Properties

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 Economic GIS (SEGIS)



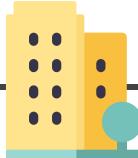
Phone Signal Data
Workday vs. Weekend
Day vs. night
(District Level)



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



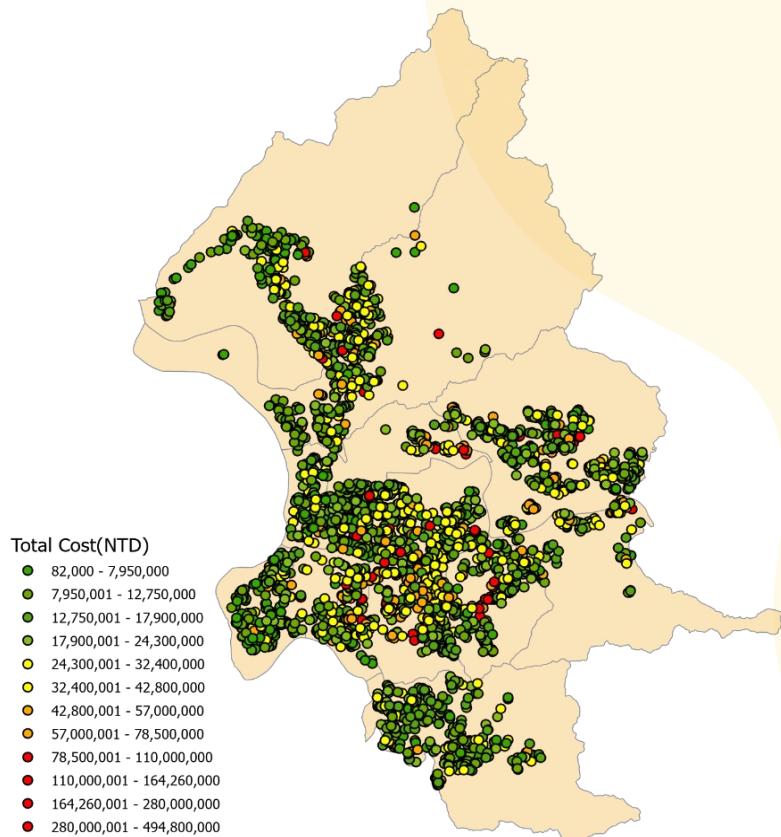
Data Transformation

Normalization of values,
One-hot encode



Ready
for
Analysis

Data Visualization: Total Price



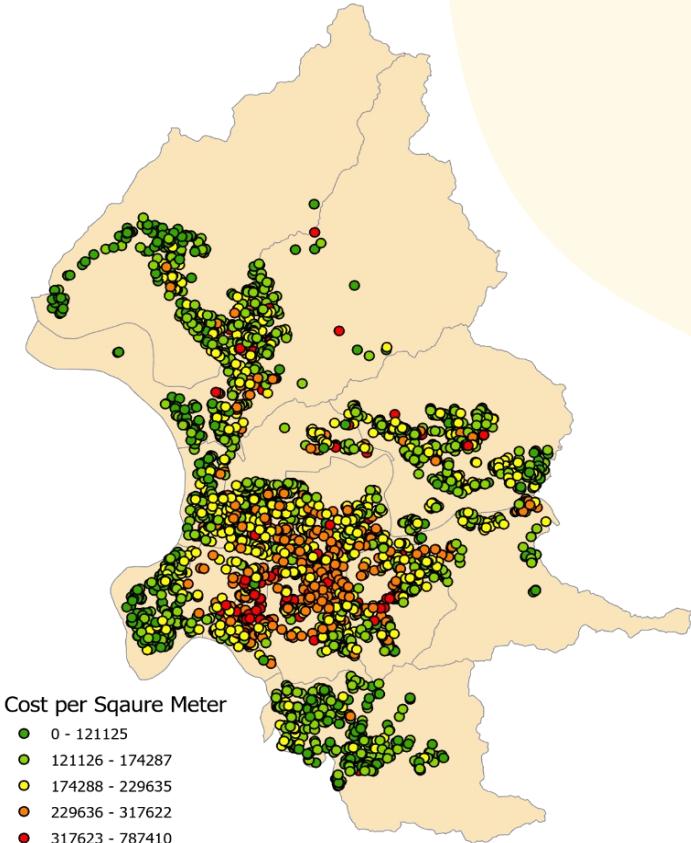
- ◻ \leq NTD \$25,000,000
(USD \$817,750)
- ◻ \geq NTD \$80,000,000
(USD \$2,616,800)

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**

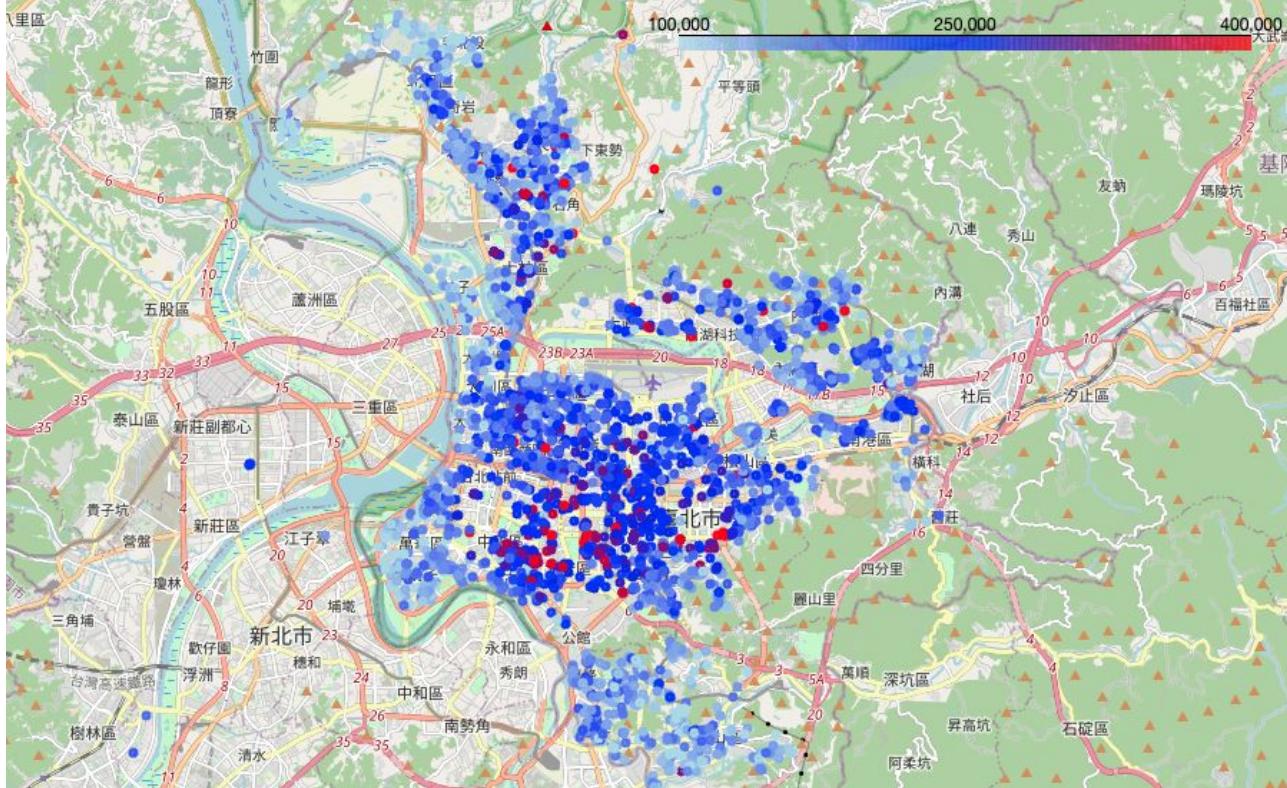


**1 kilometer
Public Transport/
Scooter Distance**



**3 kilometers
Car Distance**

Data Visualisation : Price per Square Meter (m^2)



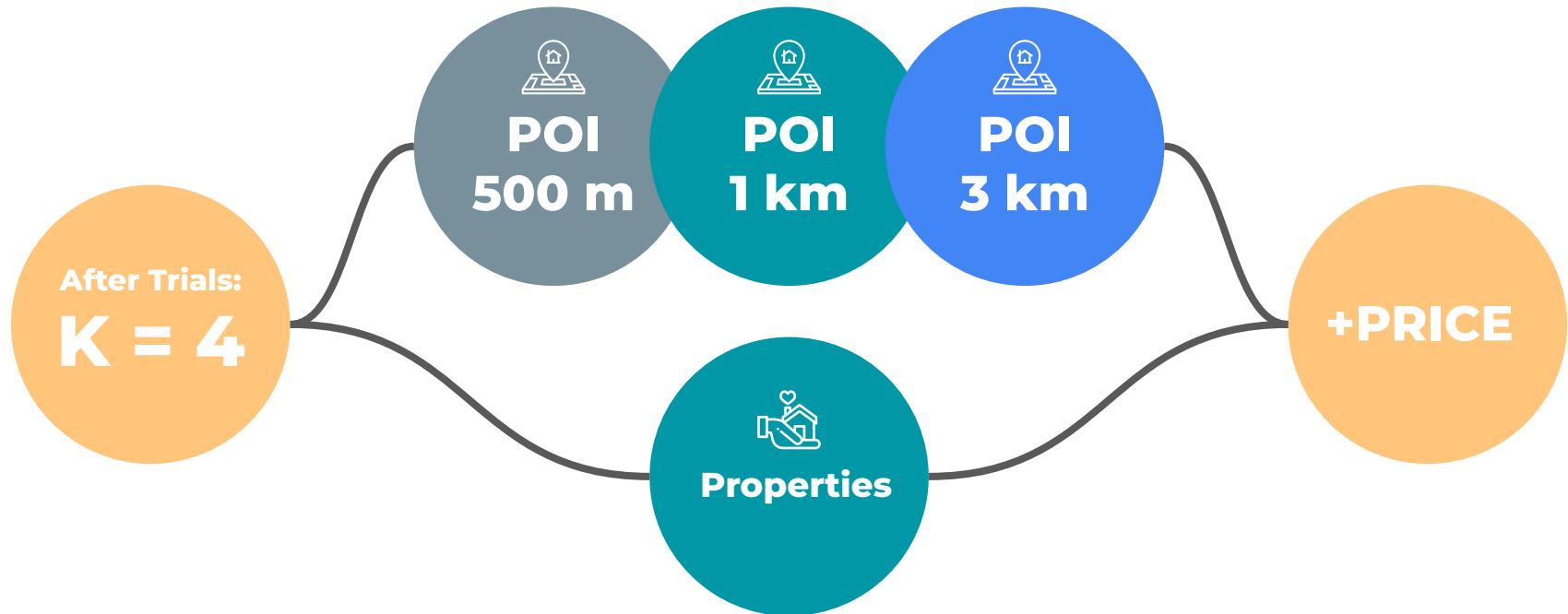
Index:

- $\leq \text{NTD } \$100,000$
(USD \$3,231)
- $\geq \text{NTD } \$400,000$
(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.

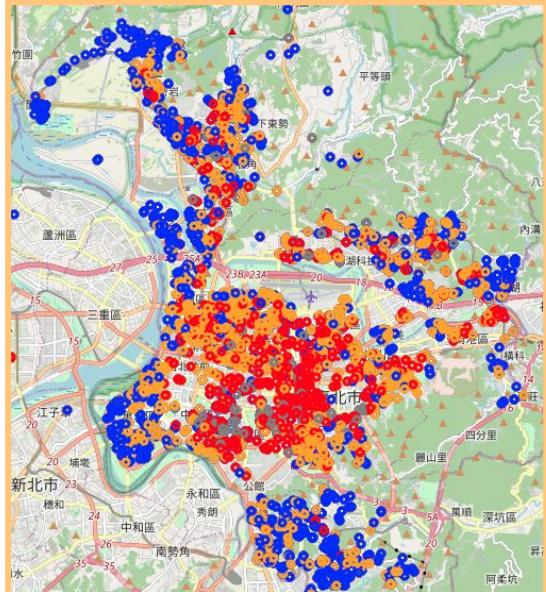
Clustering: K-Means



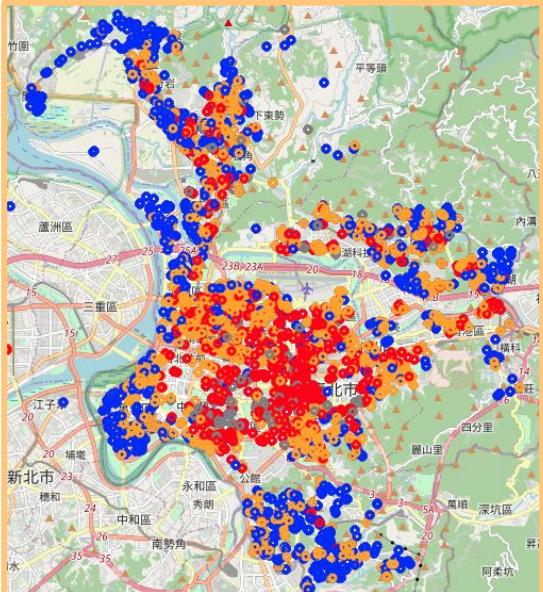
**POI vs
Properties**

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

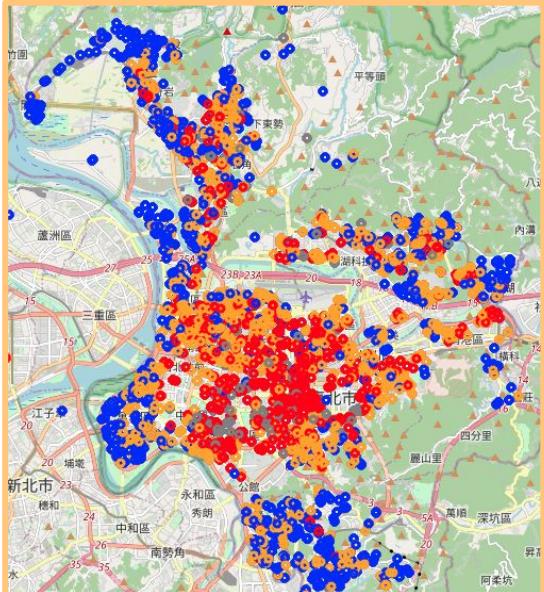
K-Means: POIs + Price



500m



1 km

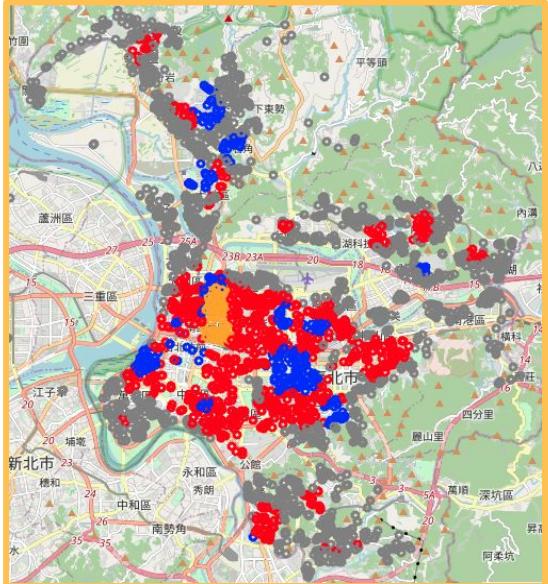


3 km

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

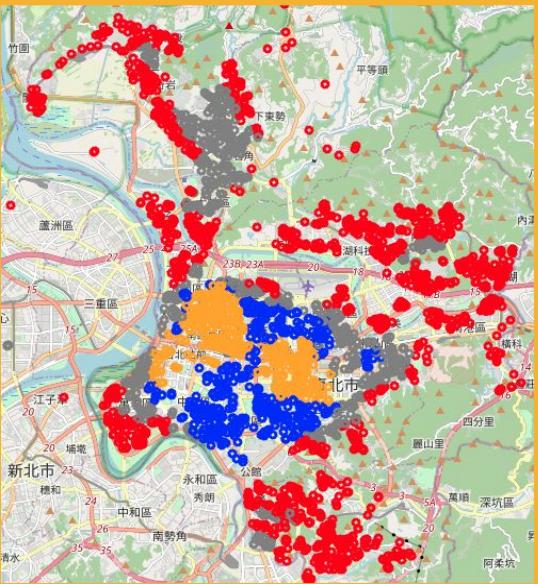
Cluster Legend:
0: Red, 1: Blue,
2: Gray, 3: Orange

K-Means: Only POIs



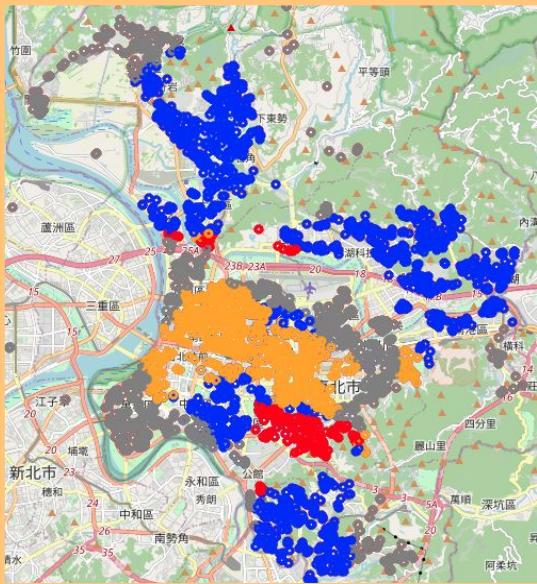
500m

There are small clusters everywhere in the city.



1 km

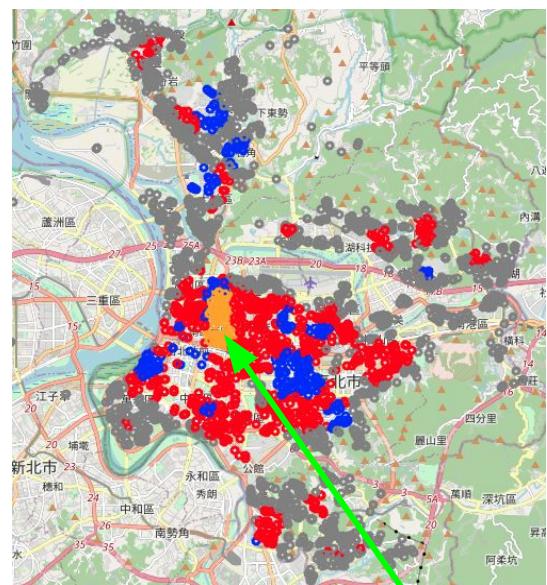
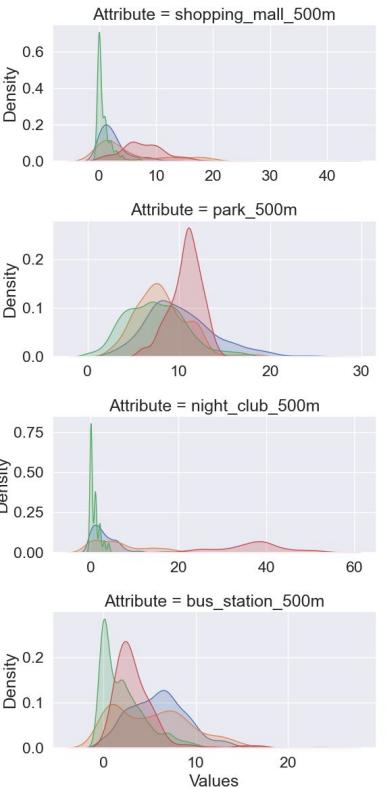
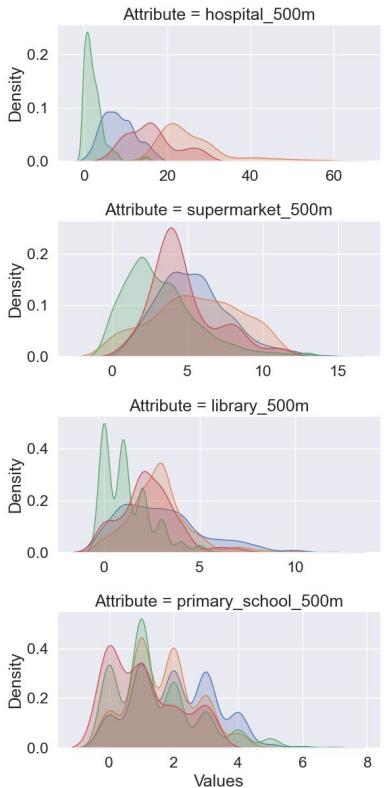
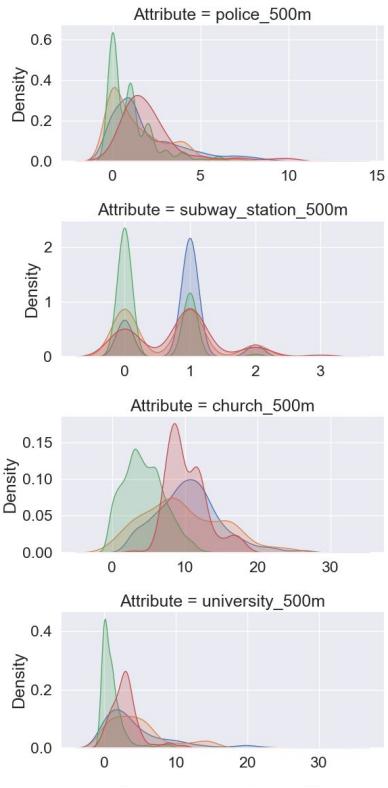
Two main cluster appear in the city, one is old town and CBD.



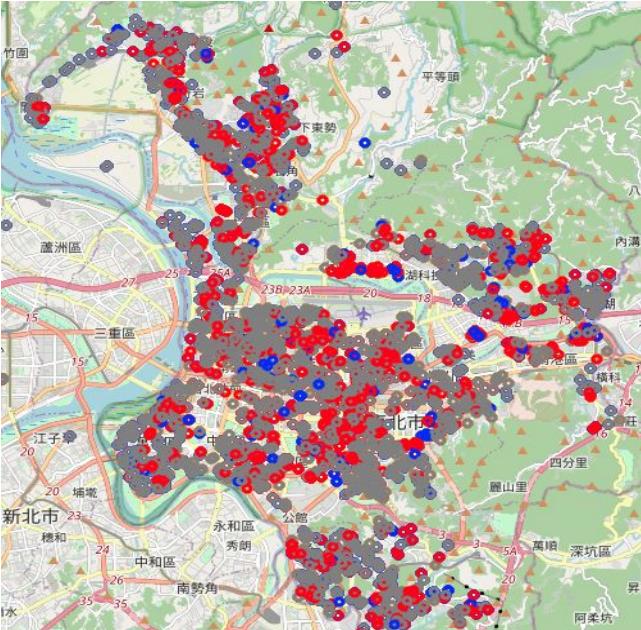
3 km

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

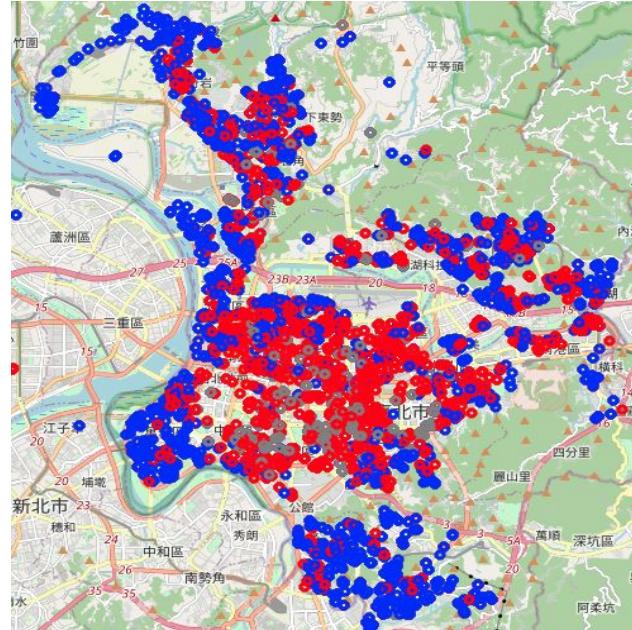
POI distributions within 500m in different clusters



K-Means: Properties / Properties + Price



Properties



Properties + Price

Findings:

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

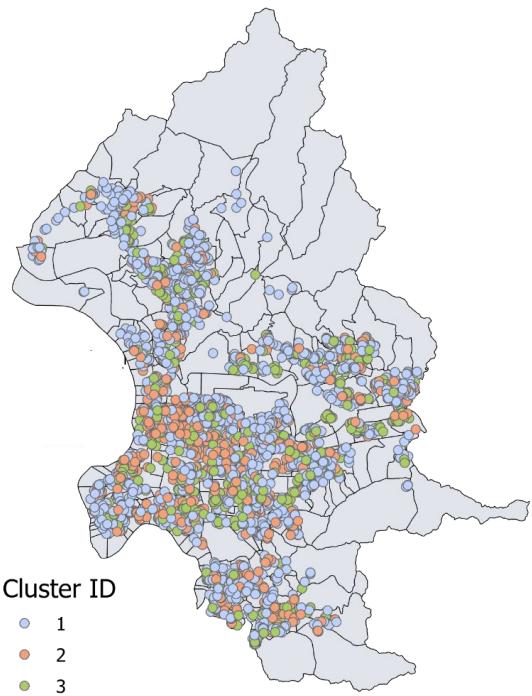
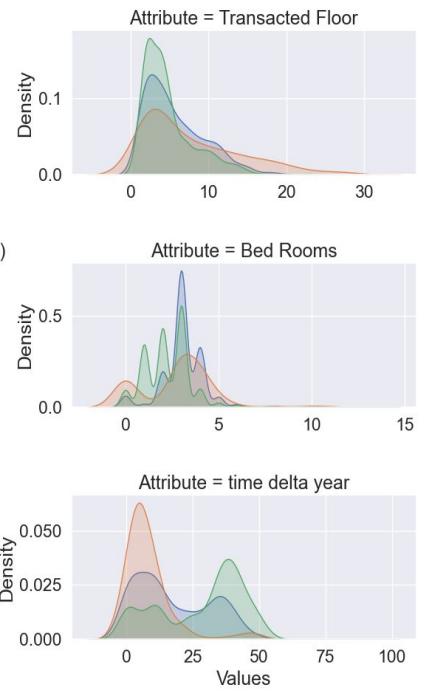
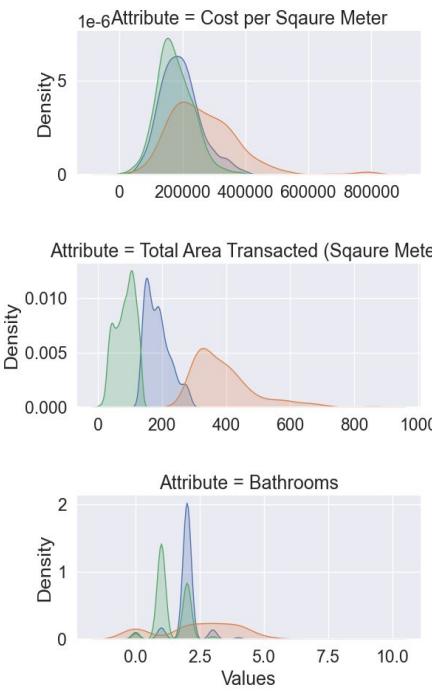
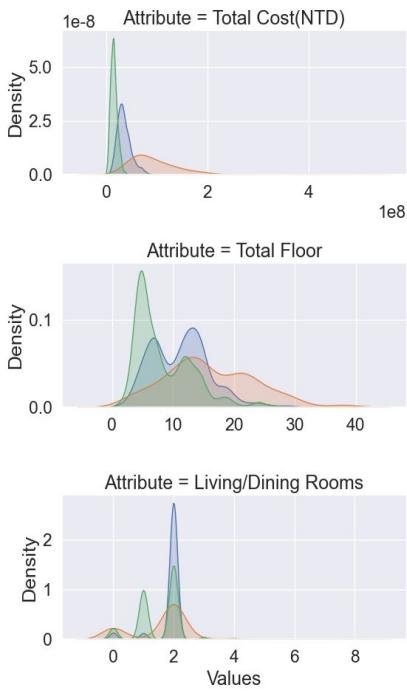
Therefore, a closer look at the clustering without price is needed.

Cluster Legend:

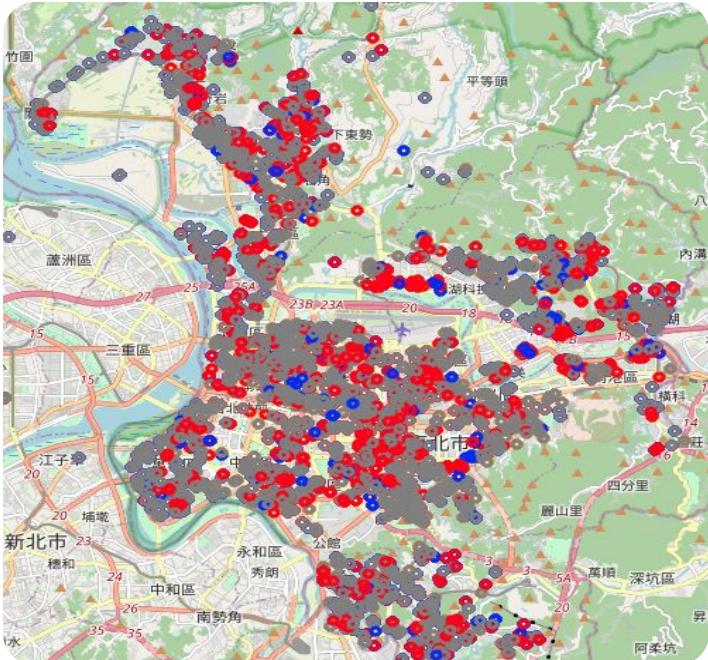
0: Red, 1: Blue, 2: Gray



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

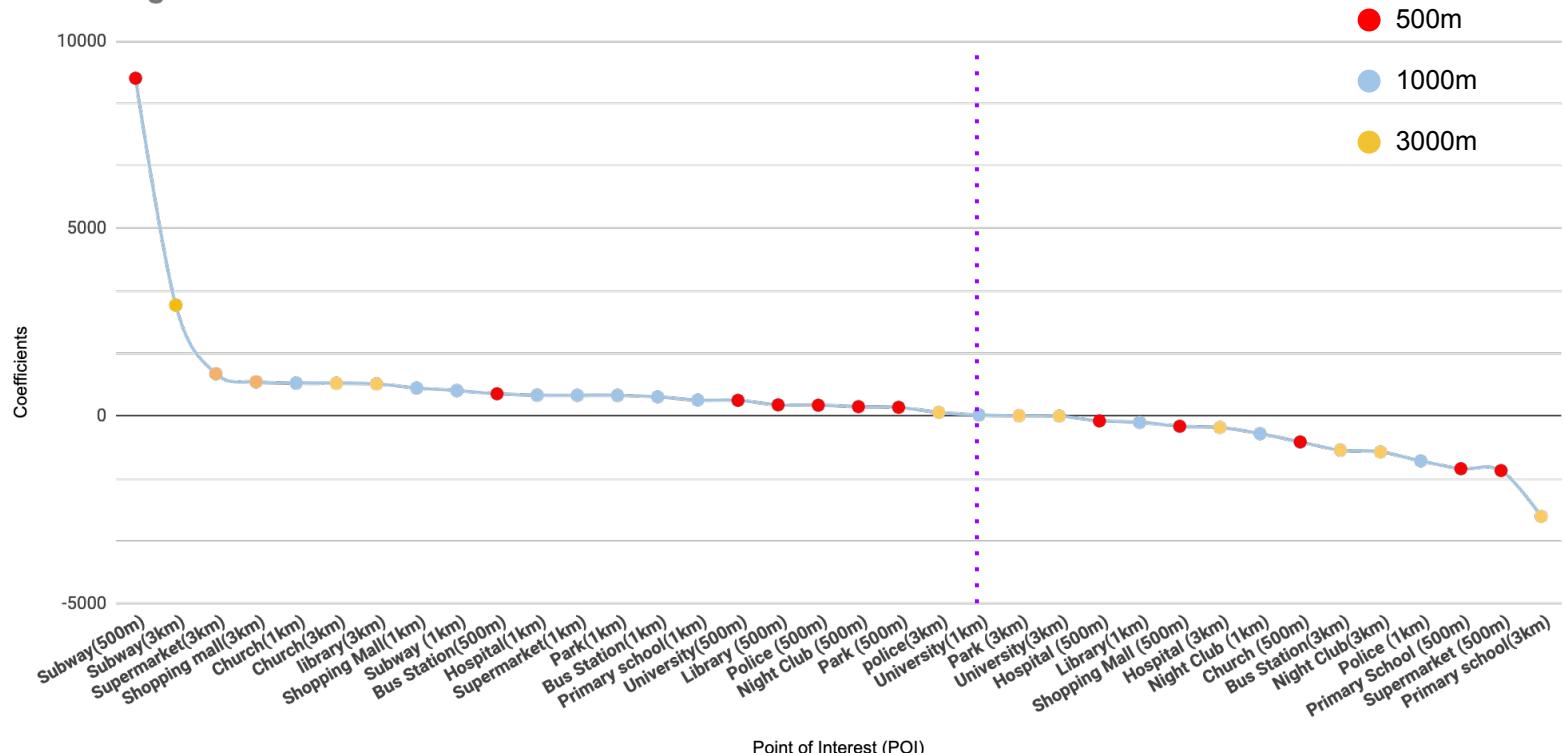


Linear Regression Model

$$Unit\ Price = \sum_i \alpha_i * F_i + bias$$

Linear Regression Model — All POI in 3 Radius

Linear Regression Coefficients for POIs in all 3 radius

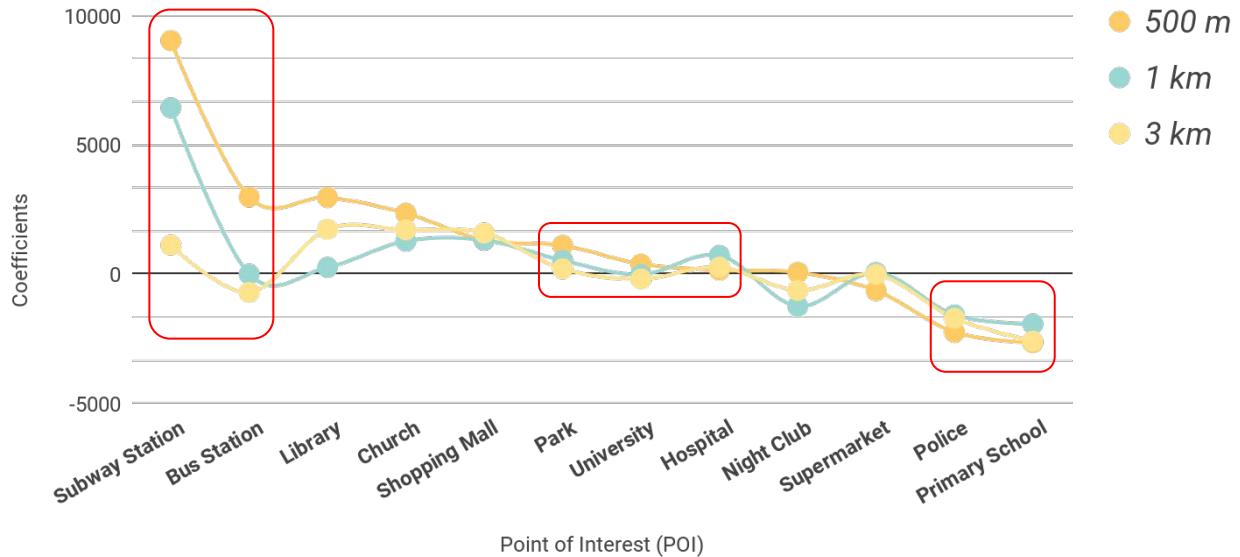


Linear Regression Model — POI

Findings:

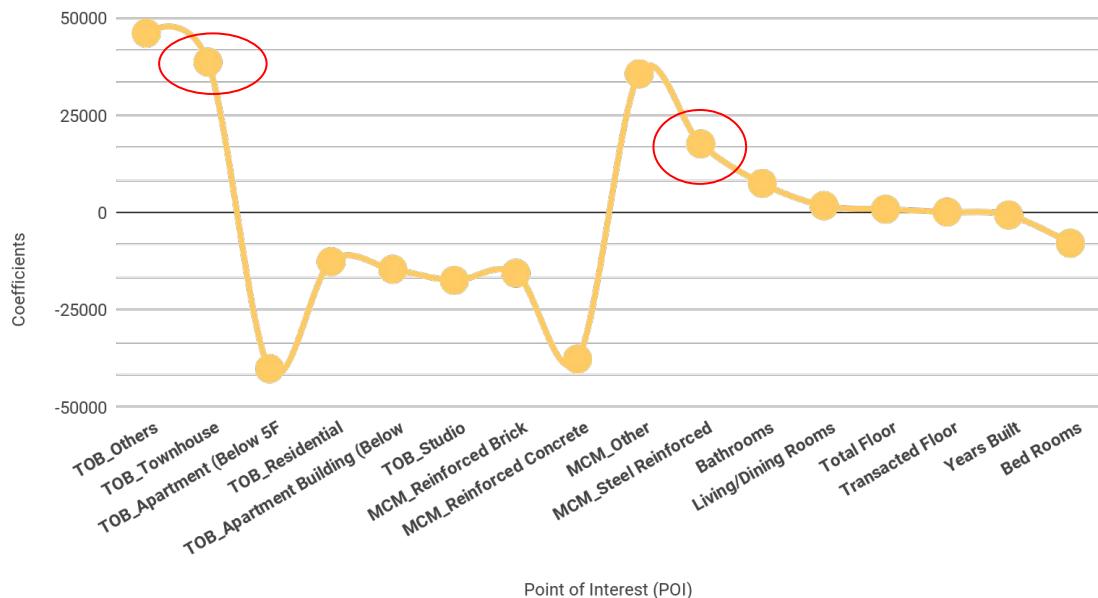
- **Subway stations and Bus stations** are important in shorter distances, and becomes less important further away.
- At 3 km, **library, church, and shopping malls** has the highest importance since these POIs may only be occasional visits.
- **Park and University's** importance would decrease as it 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

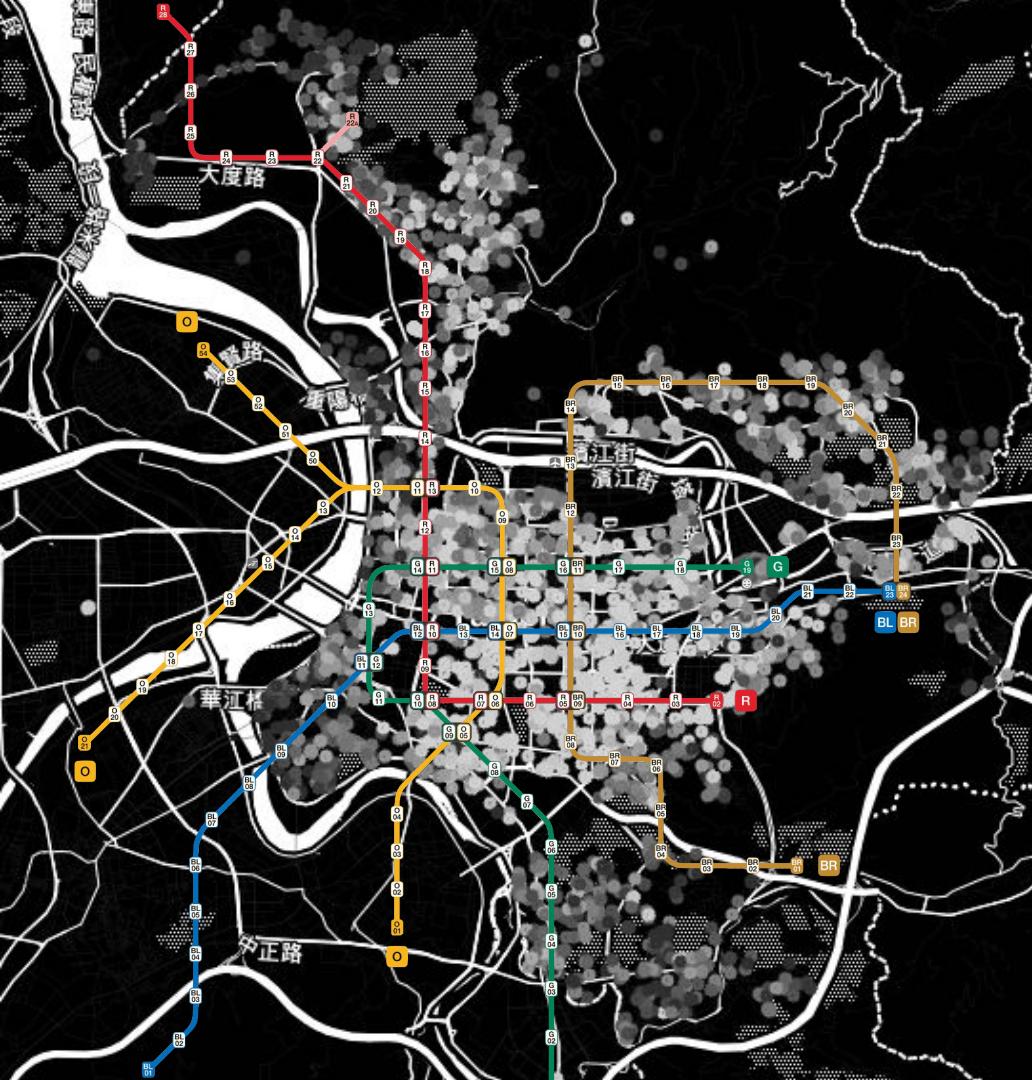


TOB → Type of Buildings

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

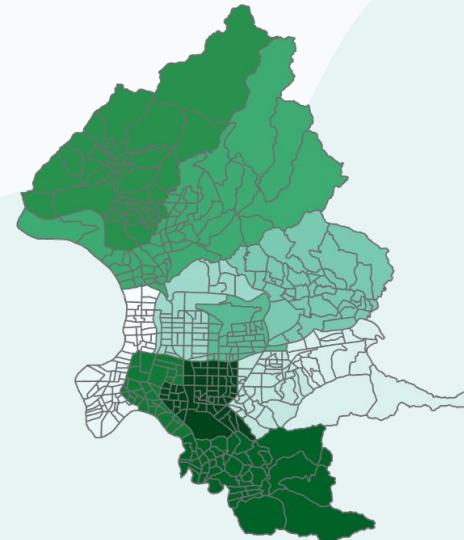
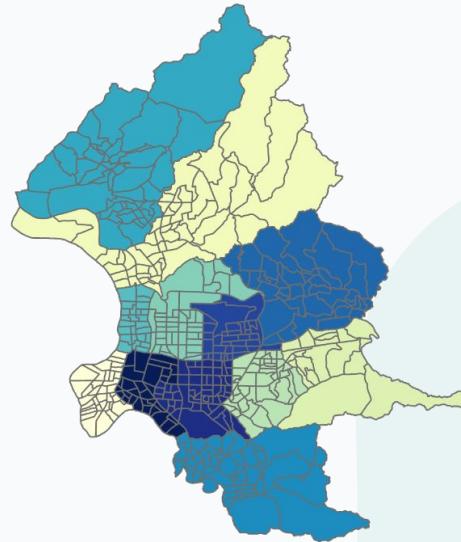
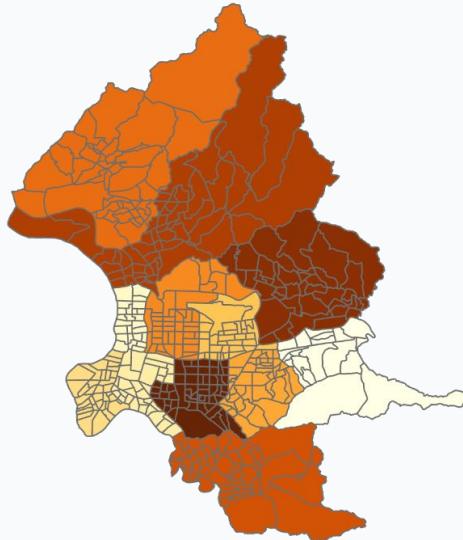
Lighter areas on this map is most expensive.



Socioeconomic Factors

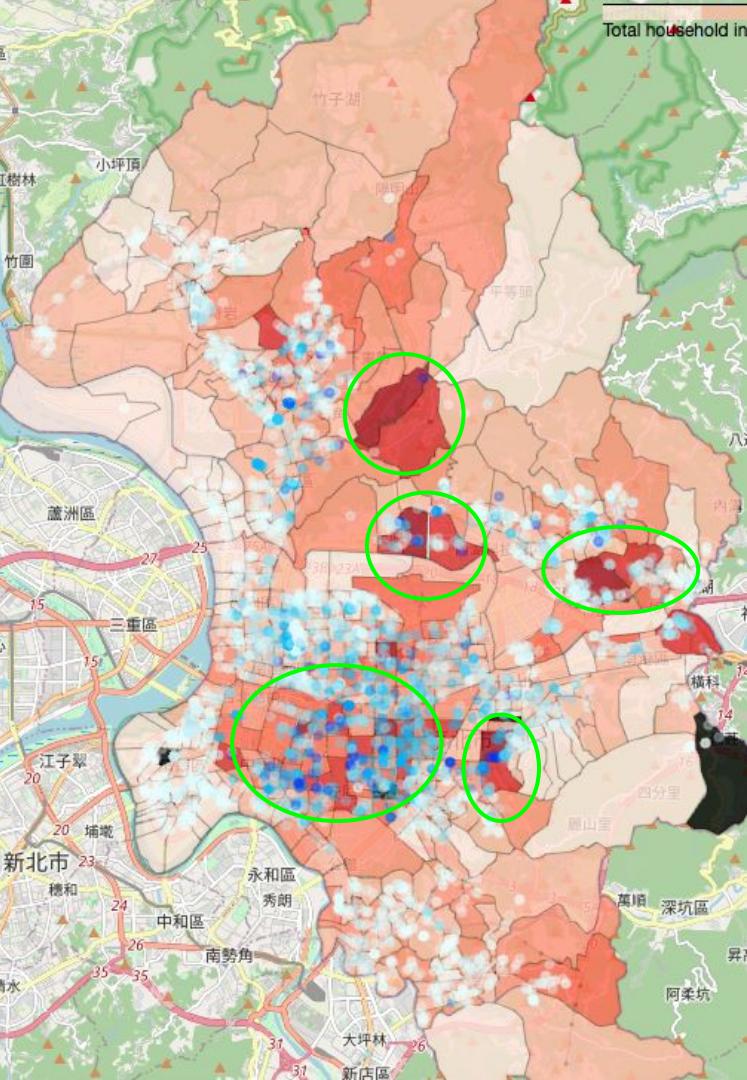
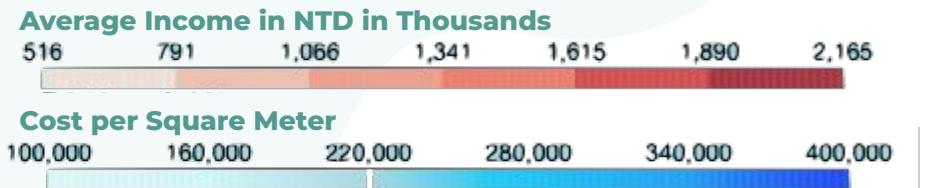


Socioeconomic Factors Overview



Average Income vs. Total House Price

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



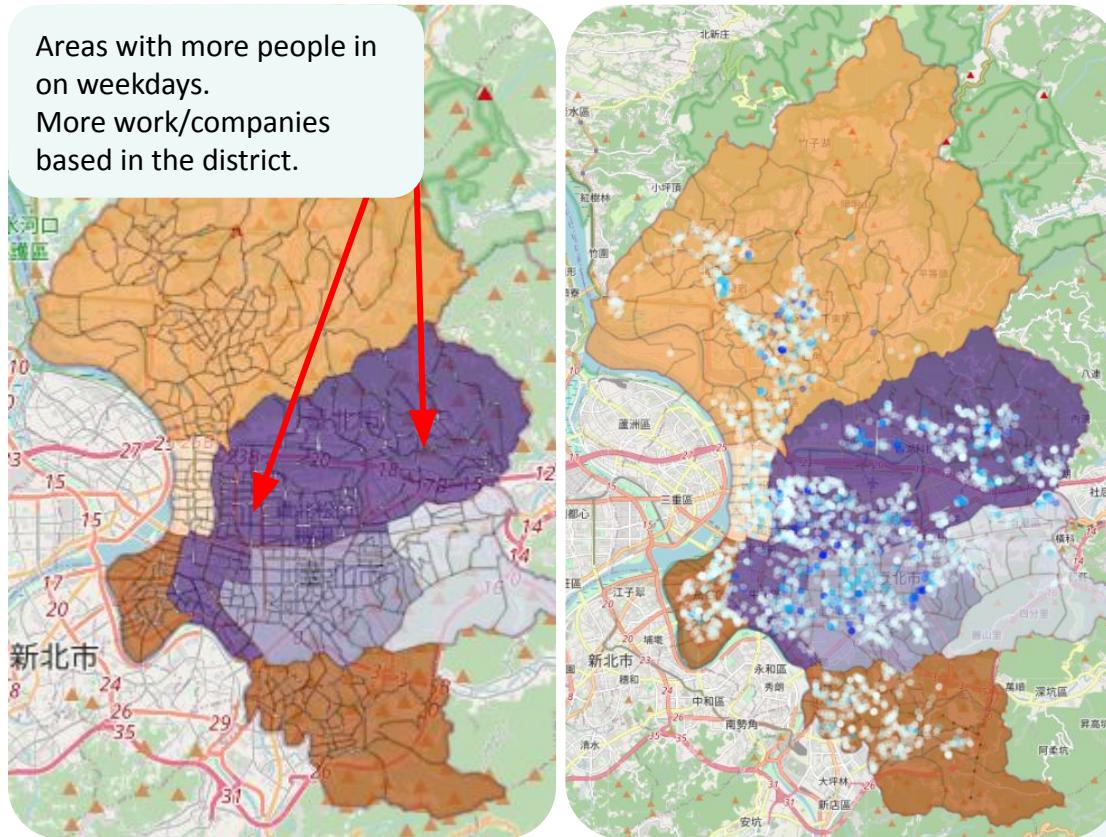
House Price v.s. Income Correlation



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.



Insights

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, live 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



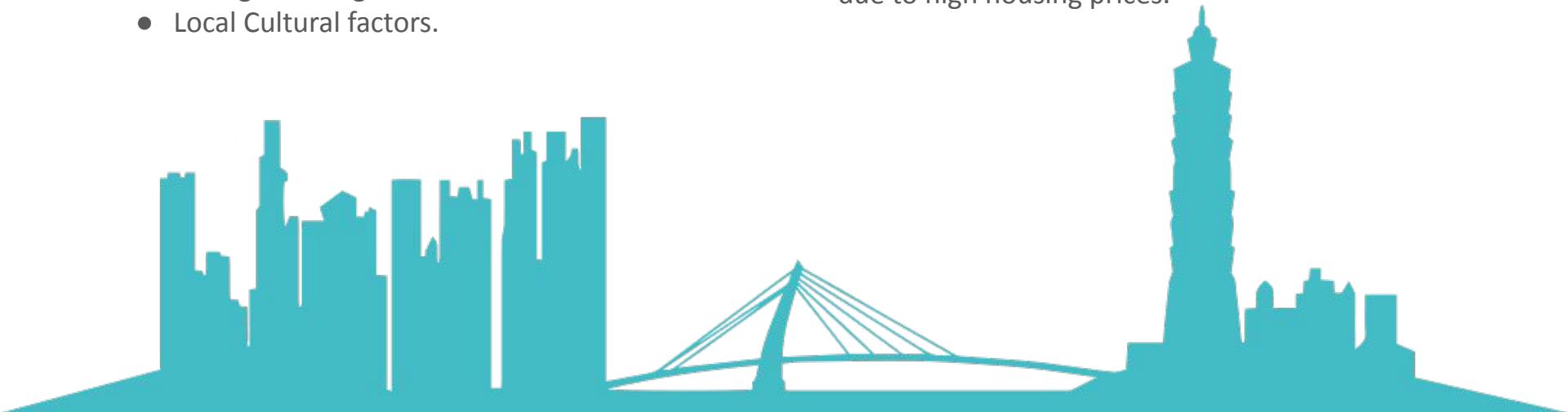
Challenges, Caveats, and Opportunities

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.

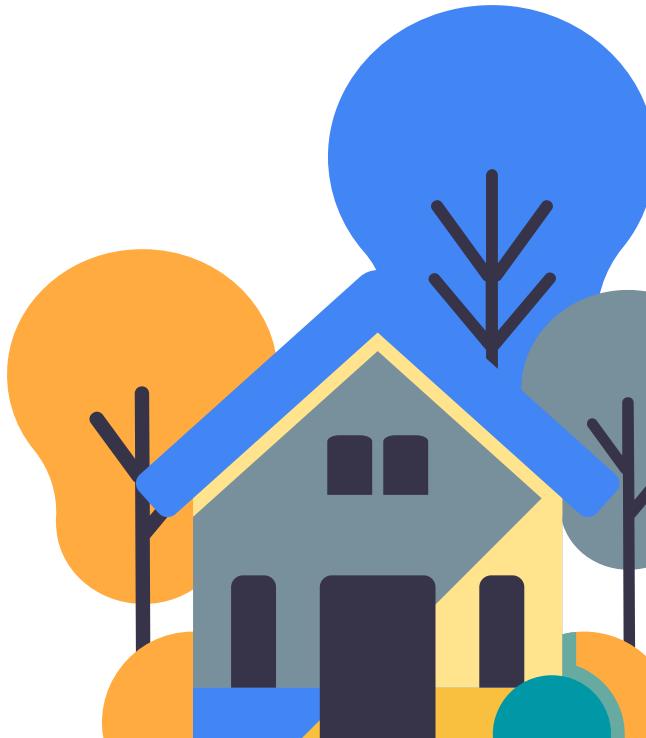
Opportunities

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



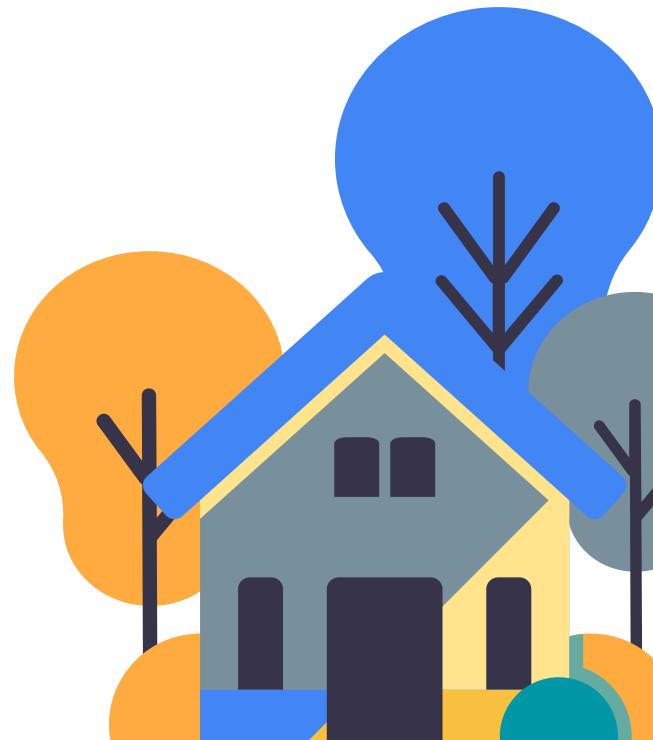
Thank you!

Questions?



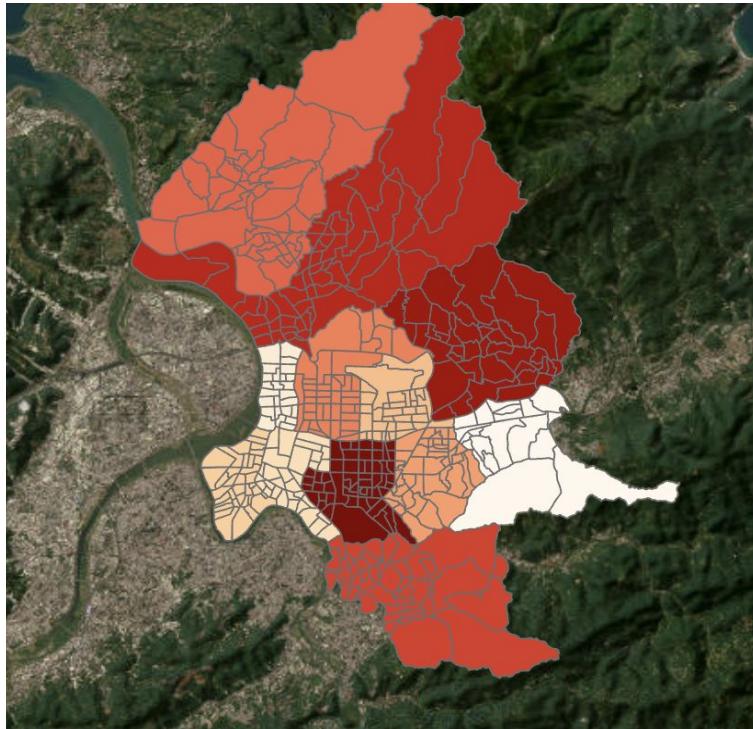
Appendix Slides

Questions?





Taipei City Background Information



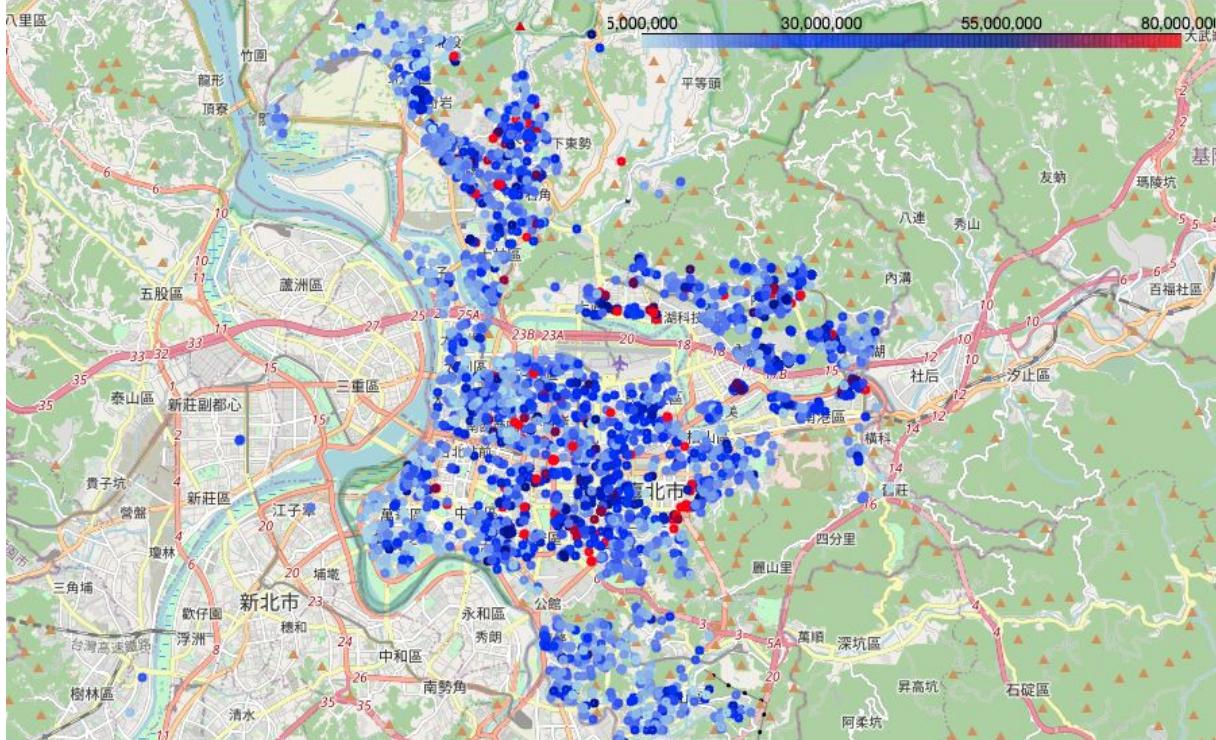
Administrative Districts:	12
Total Area:	271.80 km ²
Compared to San Francisco	121.0 km ²
Household Registered Population	2,465,610
Metropolitan Area Population:	6,894,643
Population Density:	9,071.42 / km
GDP Per capita	\$35,513
Purchasing power parities (PPP)	US\$ 760,00 in 2019
Average House Price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

Data Preview: Total Price



Index:

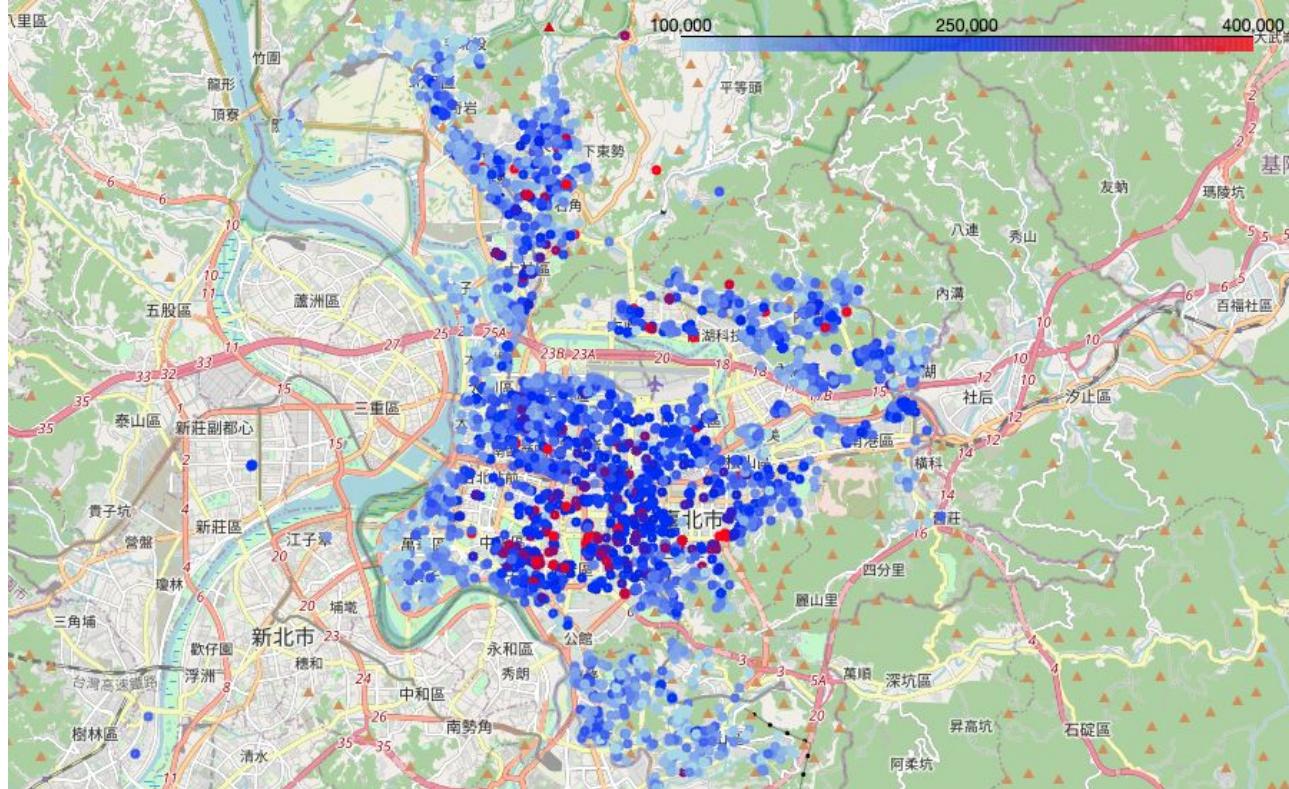
- \leq NTD \$5,000,000
(USD \$161,574)
- \geq NTD \$80,000,000
(USD \$2,585,184)

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.

Data Preview: Price per Square Meter (m²)



Index:

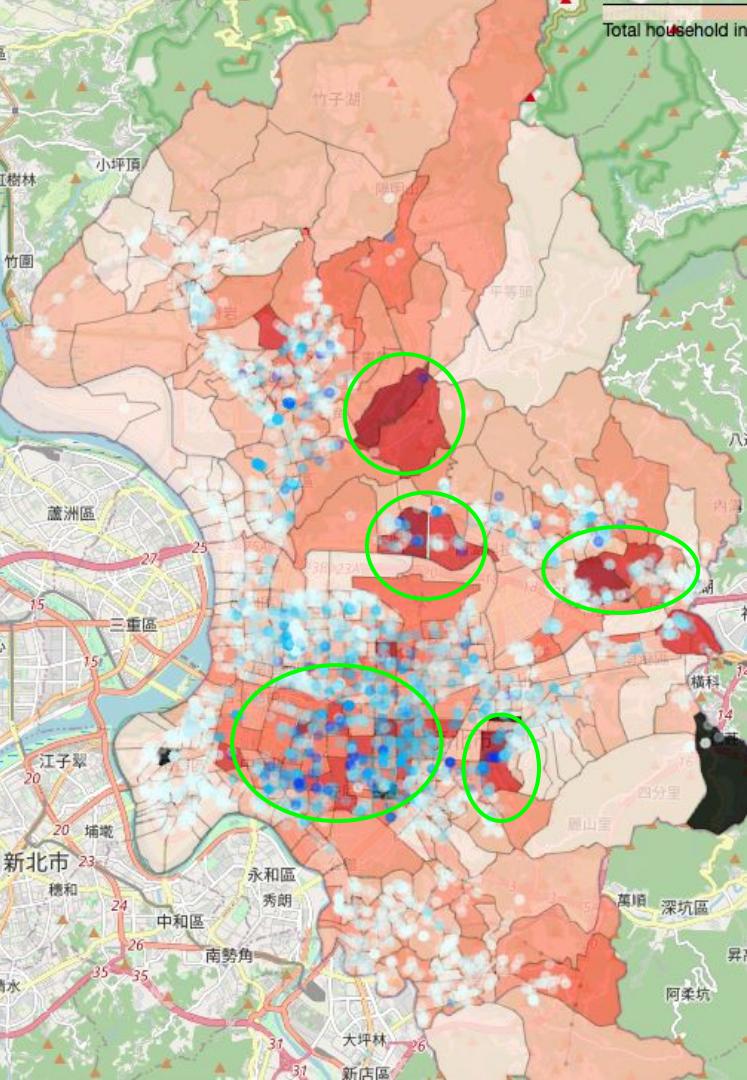
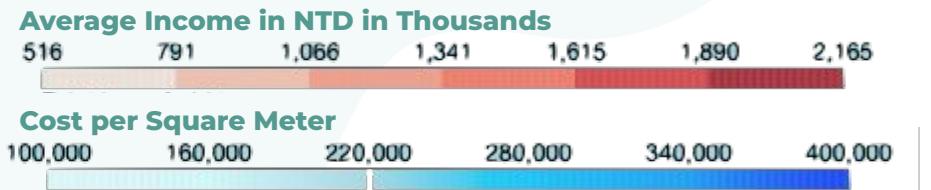
- ≤ NTD \$100,000
(USD \$3,231)
- ≥ NTD \$400,000
(USD \$12,925)

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

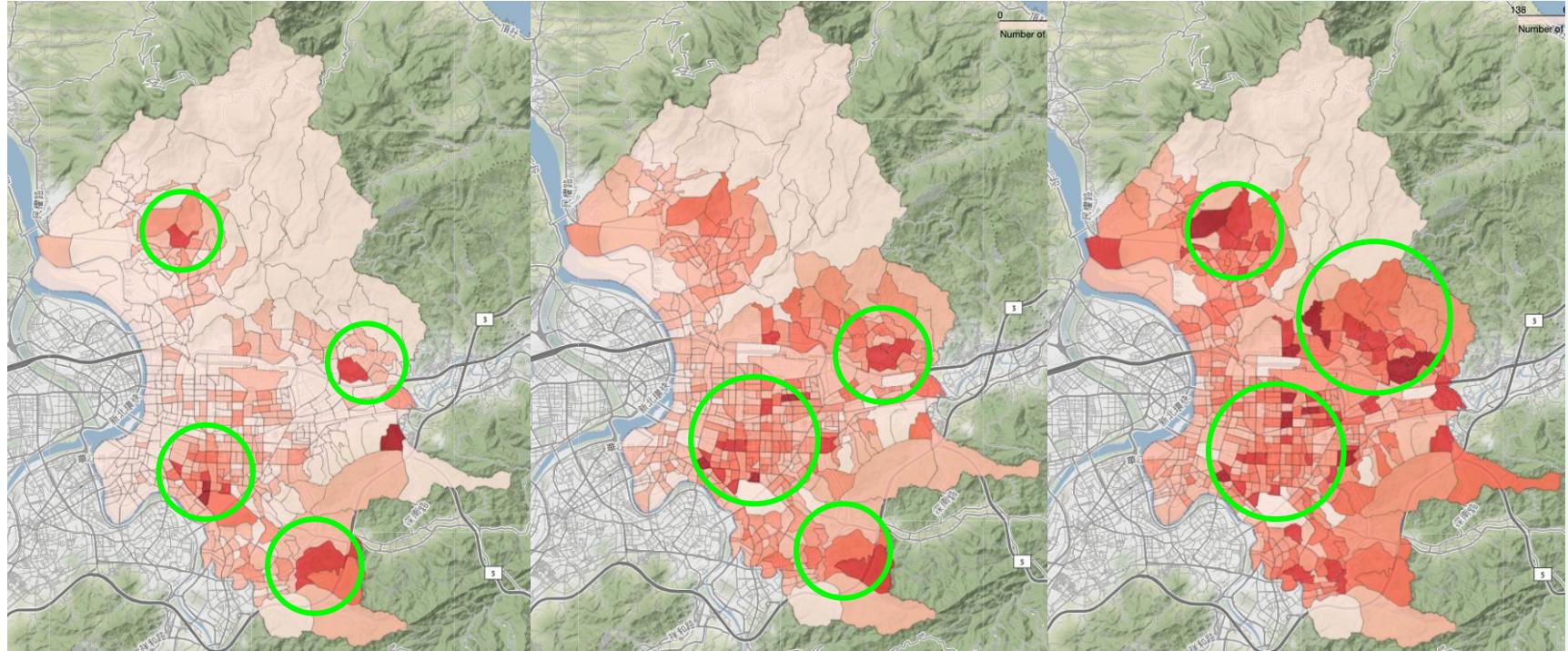
Houses with higher cost per square meter tends to land in places with higher average income.



House Price v.s. Income Correlation



Education

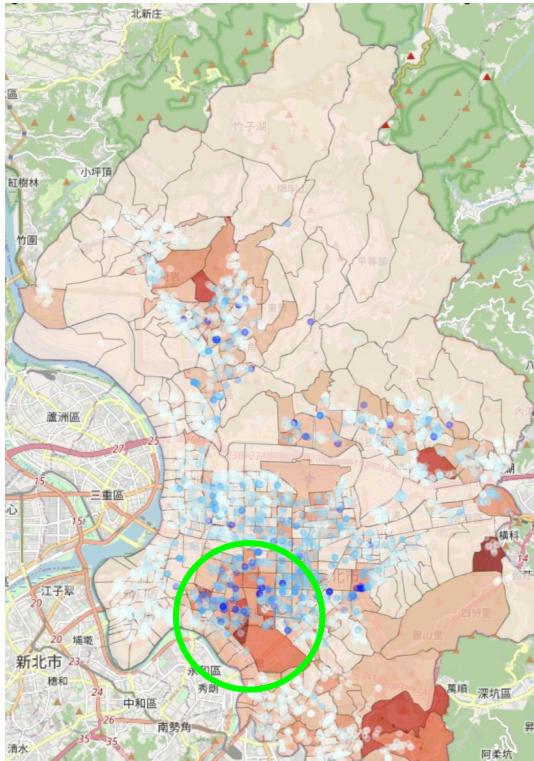


- Doctorate

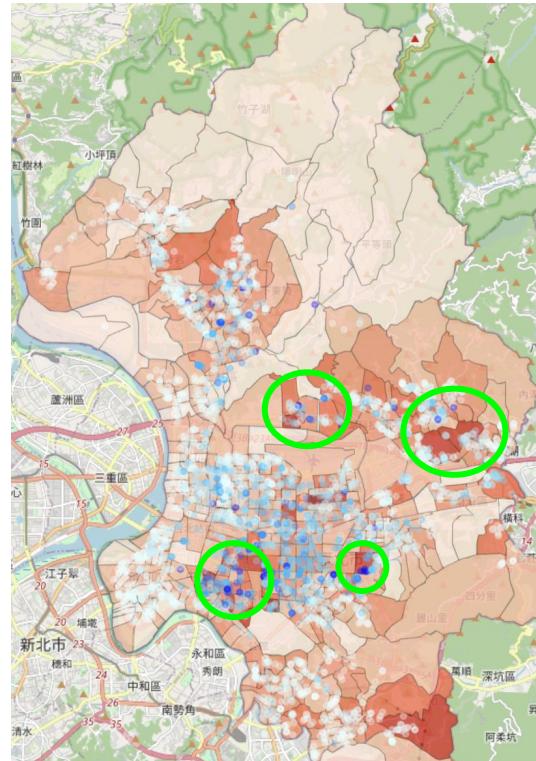
- Masters

- Undergrad

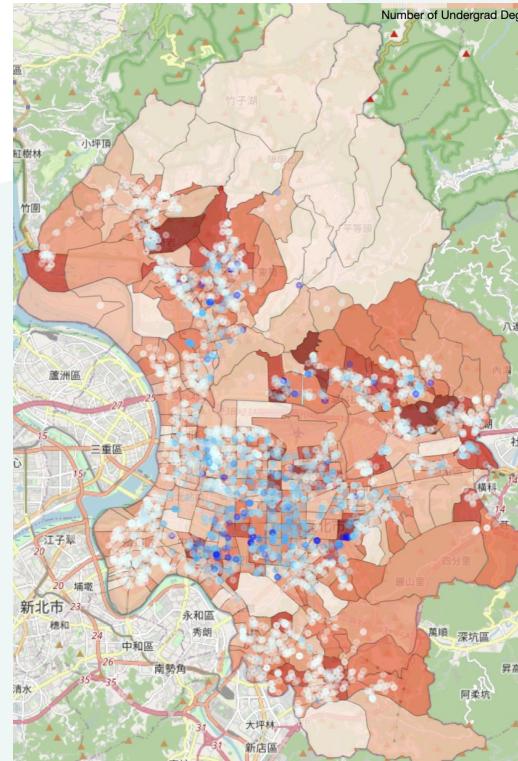
Education v.s. House Price



● Doctorate

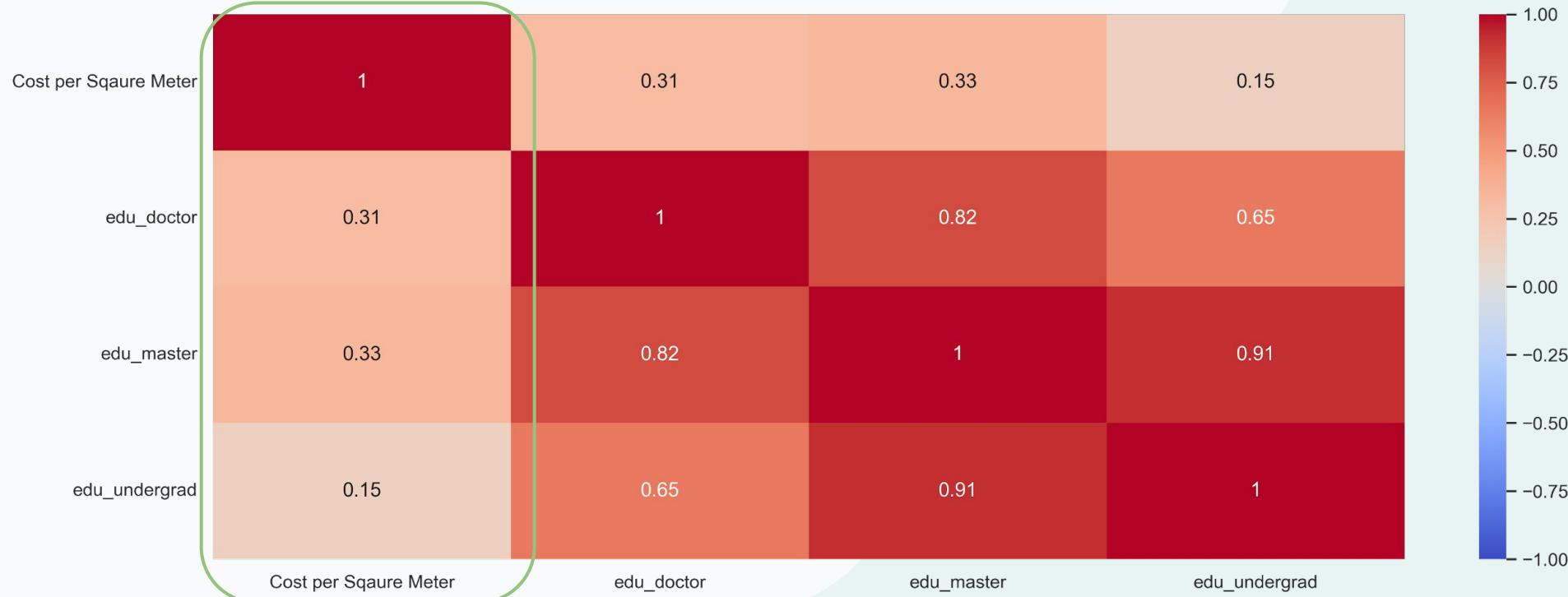


● Masters



● Undergrad

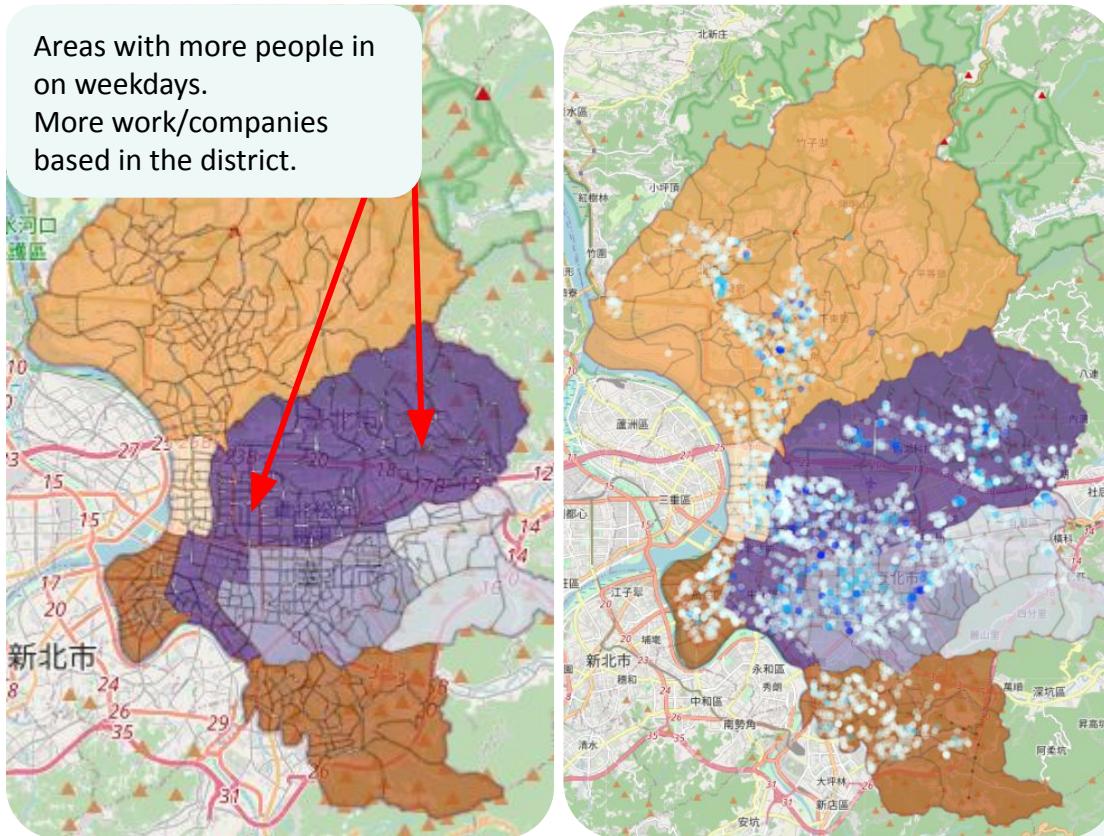
House Price v.s. Education Correlation



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

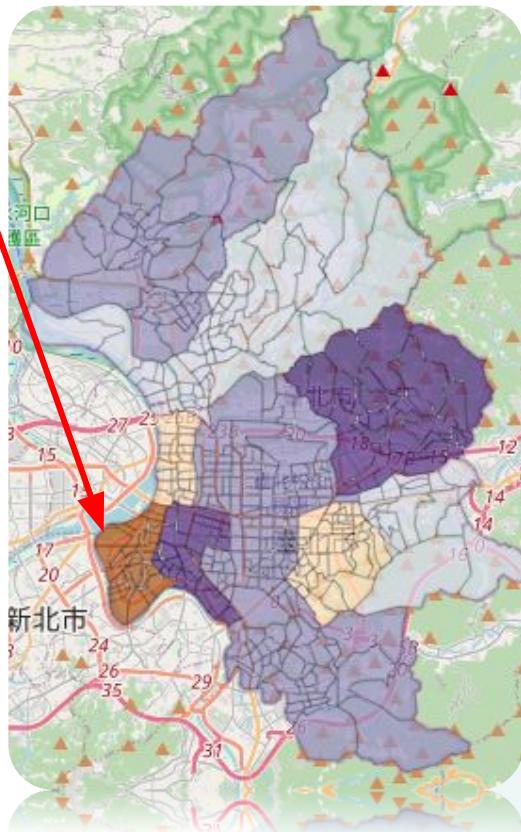
Therefore:

- > 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

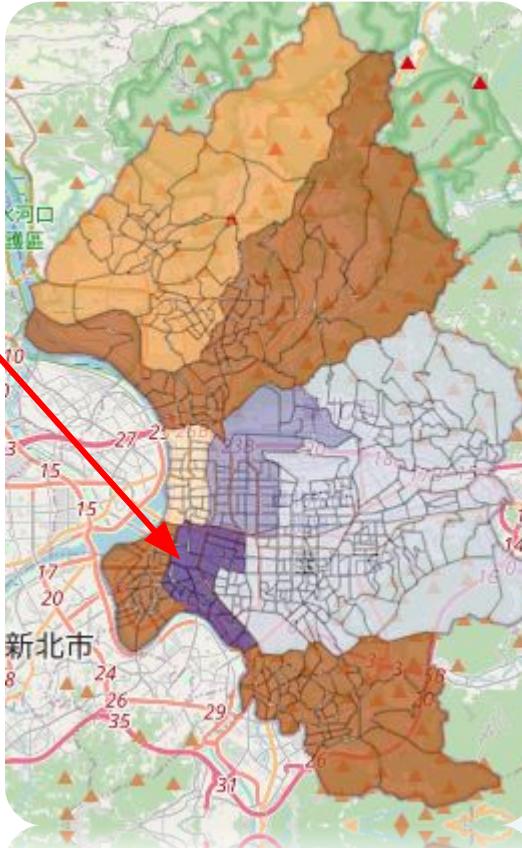
Therefore:

- >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 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.