

Seoul Kindergarten Project



MONASH
University

Gachon Univ. South Korea

KANG SEONGYEON

Contents

1. Introduction

2. Motivation

3. Data

3.1. Data source

4. Processing the data

4.1. Preprocessing

4.2. Processing

5. Analysis

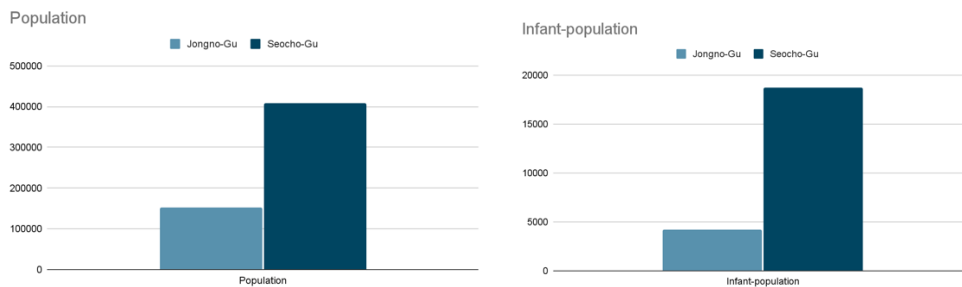
6. Conclusion

1. Introduction

Education passion in Seoul, Korea is high. This is usually revealed as a university entrance examination, but before that, fierce competition is also emerging for kindergarten entrance. In this report, I would like to analyze the factors that are factors of preference for kindergartens, propose ways to resolve such competition, and explain the expected effects.

2. Motivation

For example, there are two regions, Jongno-gu and Seocho-gu population and infant-population between the two regions are so different.



However, there is no significant difference in the number of kindergartens.

As a result, only 4 out of 10 infants in Seocho-gu go to kindergarten.



In this situation, the competition rate for kindergarten admission in Seocho-gu is bound to intensify. Nevertheless, people continue to migrate.

The reasons that I expected are below

1. People want to move to area of good infrastructure.

so I should collect the relevant data.

3. Data

Before entering the contents, let me explain the address system of Korea.

Korea>Metropolitan city/Provinces(Si/Do)>District(Gu)>Mesh block(Dong)

I Use POSTGIS for processing spatial data,

QGIS for visualization,

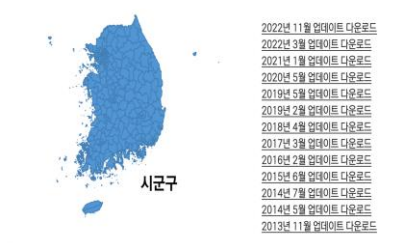
Docker for making a container and running.

3.1. Data source

1. I get the Seoul district(gu) data to analyze the difference between the districts

<http://www.gisdeveloper.co.kr/?p=2332>

website/table/visualization in qgis

[illegible]

2. Get the Korean population data

<https://data.seoul.go.kr/dataList/10718/S/2/datasetView.do>

data information

데이터 정보			
공개일자	2021.05.21.	최신수정일자	2022.04.26.
경산주기	분기(분기)	분류	복지
원본시스템		지각권자	서울특별시 디지털정책관
제공기관	서울특별시	제공부서	디지털정책관 빅데이터담당관
담당자	황재임 (02-2133-4361)		
원본형태	DB	제3자작권자	없음
라이선스	CC BY 저작표시(BY): 이용이나 변경 및 2차적 저작물의 작성을 포함한 자유이용을 허락합니다.		
관련 태그	5대법, 내국인, 등록인구, 연방법, 외국인, 한국인, 행정분별, 등록외국인, 인구, 인구수		

Tables

Enter a SQL expression to filter results

	자치구	pop	population
1	종로구	152403	152,403
2	중구	130722	130,722
3	용산구	233696	233,696
4	성동구	288584	288,584
5	광진구	350723	350,723
6	동대문구	353823	353,823
7	중랑구	390698	390,698
8	성북구	442981	442,981
9	강북구	298261	298,261
10	도봉구	314583	314,583
11	노원구	509265	509,265
12	은평구	472841	472,841
13	서대문구	318372	318,372
14	마포구	375457	375,457
15	양천구	444853	444,853
16	강서구	575456	575,456
17	구로구	417500	417,500
18	금천구	243444	243,444
19	영등포구	398392	398,392
20	동작구	392162	392,162
21	관악구	501484	501,484
22	서초구	408924	408,924
23	강남구	534885	534,885
24	송파구	665437	665,437
25	강동구	464324	464,324

**population of seoul
by district table**

Enter a SQL expression to filter results (use Ctrl+Space)

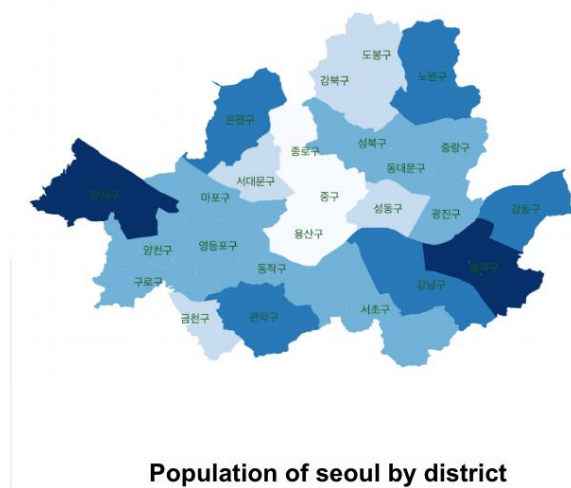
	자치구	pop	population
1	종로구	4180	4,180
2	중구	3491	3,491
3	용산구	7426	7,426
4	성동구	9781	9,781
5	광진구	10203	10,203
6	동대문구	10770	10,770
7	중랑구	11560	11,560
8	성북구	15716	15,716
9	강북구	7967	7,967
10	도봉구	9554	9,554
11	노원구	17510	17,510
12	은평구	15408	15,408
13	서대문구	10764	10,764
14	마포구	12834	12,834
15	양천구	17773	17,773
16	강서구	19893	19,893
17	구로구	14373	14,373
18	금천구	6434	6,434
19	영등포구	13360	13,360
20	동작구	12783	12,783
21	관악구	10949	10,949
22	서초구	18736	18,736
23	강남구	21298	21,298
24	송파구	27774	27,774
25	강동구	19106	19,106

**infant population of seoul
by district table**

Population of Seoul: 9,682,126

Infant-population of Seoul: 323,365

draw it as a heat map



3. get the kindergarten data

<https://download.geofabrik.de/asia.html>

Download this from the geofabrik.

Philippines	[.osm.pbf] (485 MB)	[.shp.zip]	[.osm.bz2]
Russian Federation	[.osm.pbf] (3.2 GB)	✗	[.osm.bz2]
South Korea	[.osm.pbf] (164 MB)	[.shp.zip]	[.osm.bz2]
Sri Lanka	[.osm.pbf] (107 MB)	[.shp.zip]	[.osm.bz2]
Syria	[.osm.pbf] (38.7 MB)	[.shp.zip]	[.osm.bz2]

Input the data into the database by ogr2ogr

filtered by kindergarten.

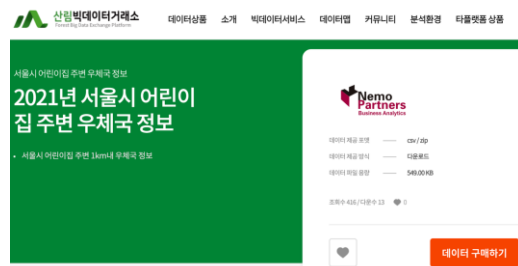
this is the result table

	oge_fid	osm_id	osm_code	fclass	name	type
1	480,806	967041988	1,500	building	과학새어린이집	kindergarten ML
2	481,038	967044676	1,500	building	논수어린이집	kindergarten ML
3	481,039	967044679	1,500	building	지구대어린이집	kindergarten ML
4	481,046	967044686	1,500	building	방울역프로토타입	kindergarten ML
5	481,064	967592004	1,500	building	세명어린이집	kindergarten ML
6	481,072	967592014	1,500	building	장원어린이집	kindergarten ML
7	482,429	968763370	1,500	building	은곡유치원	kindergarten ML
8	483,025	970538069	1,500	building	리온어린이집	kindergarten ML
9	483,213	970970893	1,500	building	한나어린이집	kindergarten ML
10	483,254	972209271	1,500	building	고양국제유치원	kindergarten ML
11	483,663	972237316	1,500	building	꽃향기유치원	kindergarten ML
12	483,937	972875286	1,500	building	[NULL]	kindergarten ML
13	484,327	973088151	1,500	building	지상어린이집	kindergarten ML
14	484,645	974881090	1,500	building	고양도 어린이집	kindergarten ML
15	484,932	975163342	1,500	building	아름다운 어린이집	kindergarten ML
16	485,234	975327739	1,500	building	과왕새 어린이집	kindergarten ML
17	485,244	975327761	1,500	building	평포동새어린이집	kindergarten ML
18	485,246	975327760	1,500	building	새벽어린이집	kindergarten ML
19	485,247	975327761	1,500	building	새벽유치원	kindergarten ML
20	485,692	975635314	1,500	building	영매유치원	kindergarten ML
21	485,705	975635335	1,500	building	죽4어린이집	kindergarten ML
22	486,086	976267676	1,500	building	[NULL]	kindergarten ML
23	486,257	976772329	1,500	building	율지어린이집	kindergarten ML
24	486,853	979921313	1,500	building	일광유치원	kindergarten ML
25	489,439	980764566	1,500	building	원화원어린이집	kindergarten ML
26	491,365	981524005	1,500	building	[NULL]	kindergarten ML
27	492,200	982373762	1,500	building	공평유치원	kindergarten ML
28	492,426	983807824	1,500	building	[NULL]	kindergarten ML
29	494,412	985066359	1,500	building	[NULL]	kindergarten ML
30	494,885	985682407	1,500	building	[NULL]	kindergarten ML
31	496,983	987025073	1,500	building	[NULL]	kindergarten ML
32	499,435	987672108	1,500	building	[NULL]	kindergarten ML

4. I want to analysis of preference factors for kindergartens in Seoul so that I get the

surrounding environment of kindergarten.

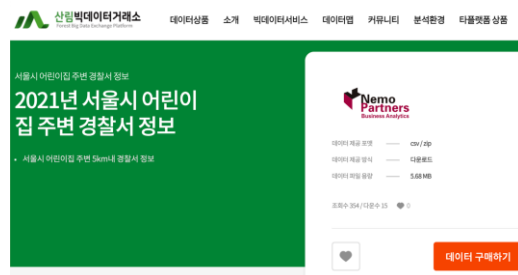
Post office: <https://www.bigdata-forest.kr/product/NMO008401>



table(403 presents, 14,590 rows)

post_idx	opg_c_fid	ctprv_n	sgng_nm	emmdn_c	emmdn_n	prc	Value
1	2	서울특별시	종로구	1101063	시작동	11110C	
2	3	서울특별시	종로구	1101063	시작동	11110C	
3	4	서울특별시	종로구	1101063	시작동	11110C	
4	5	서울특별시	종로구	1101063	시작동	11110C	
5	6	서울특별시	종로구	1101063	시작동	11110C	
6	7	서울특별시	종로구	1101063	시작동	11110C	
7	8	서울특별시	종로구	1101063	시작동	11110C	
8	9	서울특별시	종로구	1101063	시작동	11110C	
9	10	서울특별시	종로구	1101063	시작동	11110C	
10	11	서울특별시	종로구	1101063	시작동	11110C	
11	12	서울특별시	종로구	1101063	시작동	11110C	
12	13	서울특별시	종로구	1101063	시작동	11110C	
13	14	서울특별시	종로구	1101063	시작동	11110C	
14	15	서울특별시	종로구	1101063	시작동	11110C	
15	16	서울특별시	종로구	1101063	시작동	11110C	
16	17	서울특별시	종로구	1101063	시작동	11110C	
17	18	서울특별시	종로구	1101063	시작동	11110C	
18	19	서울특별시	종로구	1101063	시작동	11110C	
19	20	서울특별시	종로구	1101063	시작동	11110C	
20	21	서울특별시	종로구	1101063	시작동	11110C	
21	22	서울특별시	종로구	1101063	시작동	11110C	
22	23	서울특별시	종로구	1101063	시작동	11110C	
23	24	서울특별시	종로구	1101064	신정동	11110C	
24	25	서울특별시	종로구	1101063	시작동	11110C	
25	26	서울특별시	종로구	1101063	시작동	11110C	
26	27	서울특별시	종로구	1101063	시작동	11110C	
27	28	서울특별시	종로구	1101063	시작동	11110C	
28	29	서울특별시	종로구	1101063	시작동	11110C	
29	30	서울특별시	종로구	1101063	시작동	11110C	
30	31	서울특별시	종로구	1101063	시작동	11110C	
31	32	서울특별시	종로구	1101063	시작동	11110C	
32	33	서울특별시	종로구	1101063	시작동	11110C	
33	34	서울특별시	종로구	1101063	시작동	11110C	

Policestation: <https://www.bigdata-forest.kr/product/NMO008501>



table(274 presents, 174,718 rows)

ogc_fid	ctnr_nm	sgng_nm	amndh_c	amndh_n	prs
1	서울특별시	종로구	1101053	사직동	11110C
2	서울특별시	종로구	1101053	사직동	11110C
3	서울특별시	종로구	1101053	사직동	11110C
4	서울특별시	종로구	1101053	사직동	11110C
5	서울특별시	종로구	1101053	사직동	11110C
6	서울특별시	종로구	1101053	사직동	11110C
7	서울특별시	종로구	1101053	사직동	11110C
8	서울특별시	종로구	1101053	사직동	11110C
9	서울특별시	종로구	1101053	사직동	11110C
10	서울특별시	종로구	1101053	사직동	11110C
11	서울특별시	종로구	1101053	사직동	11110C
12	서울특별시	종로구	1101053	사직동	11110C
13	서울특별시	종로구	1101053	사직동	11110C
14	서울특별시	종로구	1101053	사직동	11110C
15	서울특별시	종로구	1101053	사직동	11110C
16	서울특별시	종로구	1101053	사직동	11110C
17	서울특별시	종로구	1101053	사직동	11110C
18	서울특별시	종로구	1101053	사직동	11110C
19	서울특별시	종로구	1101053	사직동	11110C
20	서울특별시	종로구	1101053	사직동	11110C
21	서울특별시	종로구	1101053	사직동	11110C
22	서울특별시	종로구	1101053	사직동	11110C
23	서울특별시	종로구	1101053	사직동	11110C
24	서울특별시	종로구	1101053	사직동	11110C
25	서울특별시	종로구	1101053	사직동	11110C
26	서울특별시	종로구	1101053	사직동	11110C
27	서울특별시	종로구	1101053	사직동	11110C
28	서울특별시	종로구	1101053	사직동	11110C
29	서울특별시	종로구	1101053	사직동	11110C
30	서울특별시	종로구	1101053	사직동	11110C
31	서울특별시	종로구	1101053	사직동	11110C
32	서울특별시	종로구	1101053	사직동	11110C

Mart: <https://www.bigdata-forest.kr/product/NMO008301>

산림빅데이터거래소
Forest Big Data Exchange Platform

데이터상품 소개 빅데이터서비스 데이터맵 커뮤니티 분석환경 타플렛상품

서울시 어린이집 주변 대형마트 정보

2021년 서울시 어린이집 주변 대형마트 정보

서울시 어린이집 주변 1km내 대형마트 정보

Nemo Partners Business Analytics

데이터 제공 포맷: csv/xlsx
데이터 제공 방식: 다운로드
데이터 파일 용량: 380.00 KB

조회수 381 / 다운로드 13

데이터 구매하기

table(70 presents, 3,373 rows)

ogc_fid	ctnr_nm	sgng_nm	amndh_c	amndh_n	prs
1	서울특별시	종로구	1101061	종로1-2-3-4가동	11110C
2	서울특별시	강남구	1102055	명동	11305S
3	서울특별시	강남구	1102055	명동	11480S
4	서울특별시	강서구	1102055	명동	11500S
5	서울특별시	종로구	1101067	종로1동	11110C
6	서울특별시	종로구	1101067	종로2동	11280S
7	서울특별시	강북구	1108071	동북1동	11290S
8	서울특별시	강북구	1108071	동북1동	11290S
9	서울특별시	종로구	1101067	종로1동	11110C
10	서울특별시	종로구	1101067	종로2동	11280S
11	서울특별시	강북구	1108071	동북1동	11290S
12	서울특별시	강북구	1108071	동북2동	11320S
13	서울특별시	강북구	1108071	동북3동	11360S
14	서울특별시	강북구	1108071	동북4동	11390S
15	서울특별시	강북구	1108071	동북5동	11445S
16	서울특별시	강북구	1108071	동북6동	11490S
17	서울특별시	강북구	1108071	동북7동	11545S
18	서울특별시	강북구	1108071	동북8동	11590S
19	서울특별시	강북구	1108071	동북9동	11640S
20	서울특별시	강북구	1108071	동북10동	11690S
21	서울특별시	강북구	1108071	동북11동	11740S
22	서울특별시	강북구	1108071	동북12동	11790S
23	서울특별시	강북구	1108071	동북13동	11840S
24	서울특별시	강북구	1108071	동북14동	11890S
25	서울특별시	강북구	1108071	동북15동	11940S
26	서울특별시	강북구	1108071	동북16동	11990S
27	서울특별시	강북구	1108071	동북17동	12040S
28	서울특별시	강북구	1108071	동북18동	12090S
29	서울특별시	강북구	1108071	동북19동	12140S
30	서울특별시	강북구	1108071	동북20동	12190S
31	서울특별시	강북구	1108071	동북21동	12240S
32	서울특별시	강북구	1108071	동북22동	12290S

Park: <https://www.bigdata-forest.kr/product/NMO008201>

12 Enter a SQL query to filter your data by date range									
	orig_id	city_nm	orig_nm	amend_cd	amend_nm	prj	value		
1	3	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
2	4	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
3	5	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
4	6	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
5	7	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
6	8	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
7	9	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
8	10	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
9	11	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
10	12	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
11	13	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
12	14	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
13	15	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
14	16	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
15	17	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
16	18	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
17	19	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
18	20	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
19	21	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
20	22	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
21	23	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
22	24	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
23	25	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
24	26	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
25	27	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
26	28	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
27	29	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
28	30	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
29	31	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
30	32	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
31	33	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
32	34	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			
33	35	札幌市厚別区	厚別区庁舎	1101063	庁舎新築	1110106			

A map of the Iberian Peninsula showing the distribution of sampling points. The map is divided into administrative regions by blue lines. Sampling points are represented by colored dots (red, green, yellow, and blue) scattered across the landmass, with a high concentration in the central and eastern parts of the peninsula.

5. Get the percentage of high-income earners by districts in Seoul

Seoul Research Institute, 2014, Seoul Survey,

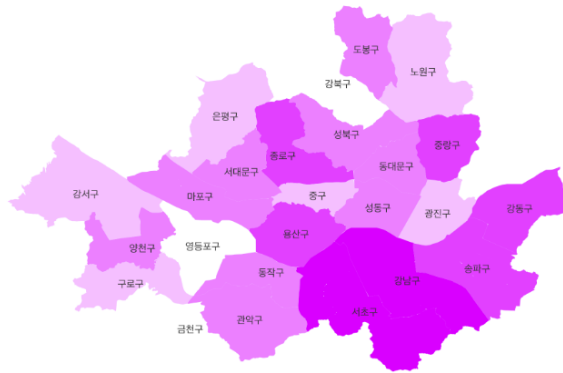
https://www.si.re.kr/si_download/56731/18179,

<https://v.daum.net/v/20220717060141744>

Result table

	ABC 자치구 ▼	123 percentage ▼
1	종로구	28
2	중구	16.5
3	용산구	27
4	성동구	23.2
5	광진구	17.4
6	동대문구	23.6
7	종랑구	31.6
8	성북구	22.6
9	강북구	10.4
10	도봉구	25.5
11	노원구	17.6
12	은평구	18.9
13	서대문구	22
14	마포구	21.2
15	양천구	26.9
16	강서구	15.2
17	구로구	15.9
18	금천구	11.3
19	영등포구	7
20	동작구	21
21	관악구	21.3
22	서초구	40.2
23	강남구	37.1
24	송파구	33.4
25	강동구	27.4

Visualization



percentage of high-income earners by districts(Gu) in Seoul

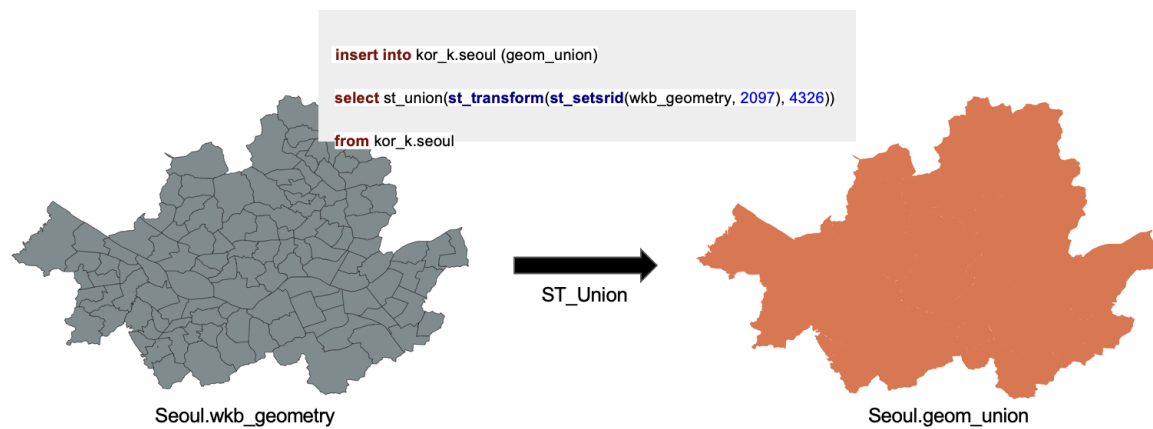
the darker the color, the richer the area

4. Processing the data

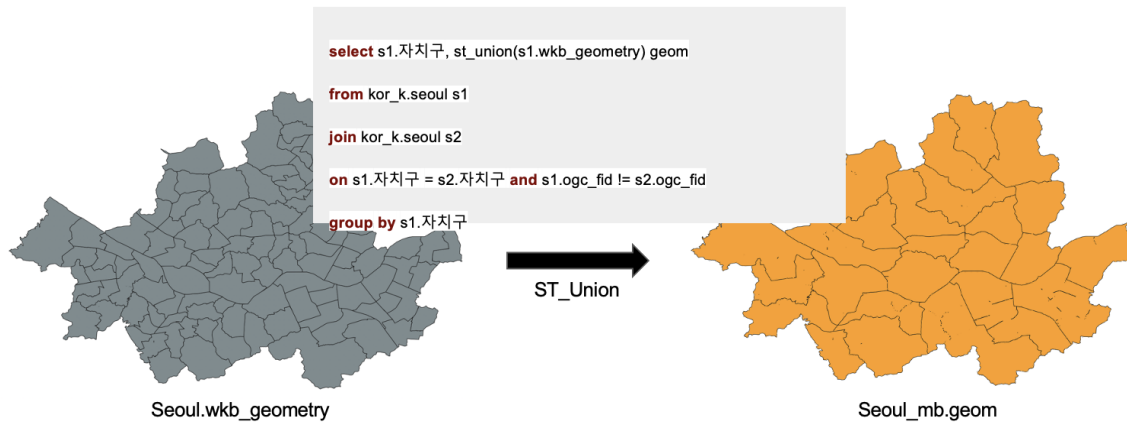
4.1. Preprocessing

I need to filtering the kindergartens only in seoul.

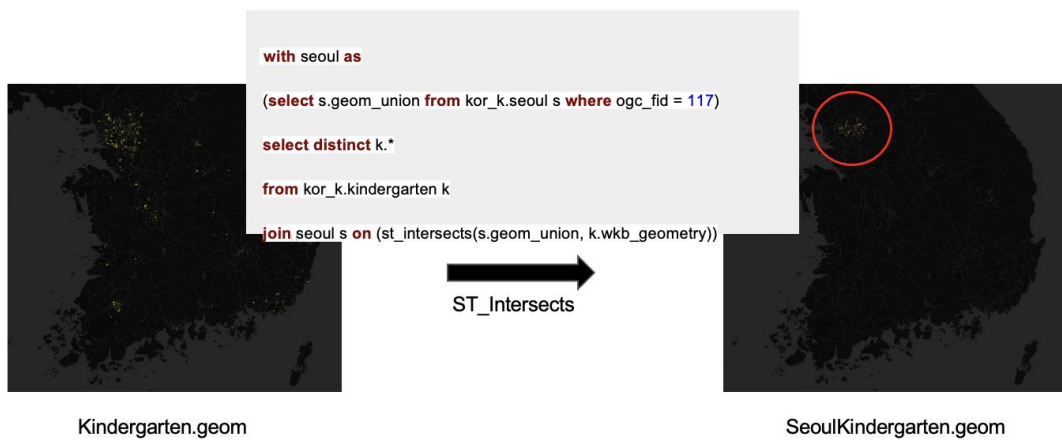
step 1: Union area for whole Seoul



step 2: Union area for each Seoul district

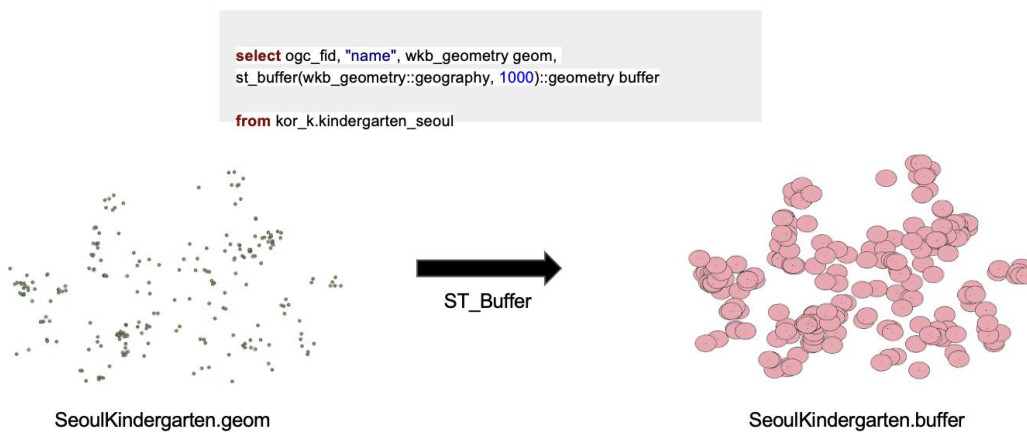


step 3: Filter kindergartens in Seoul



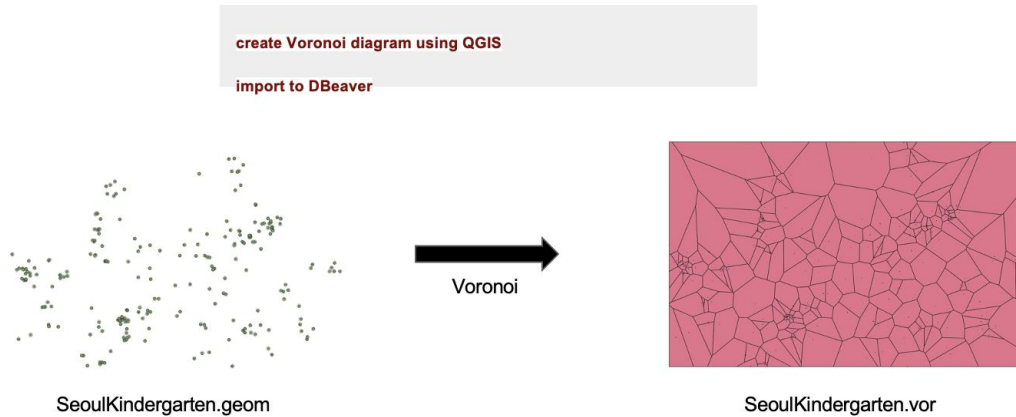
Create the radius 1.0 km for each kindergarten.

I will use this later for searching the surrounding environment of each kindergarten.



Create the Voronoi diagram for each kindergarten.

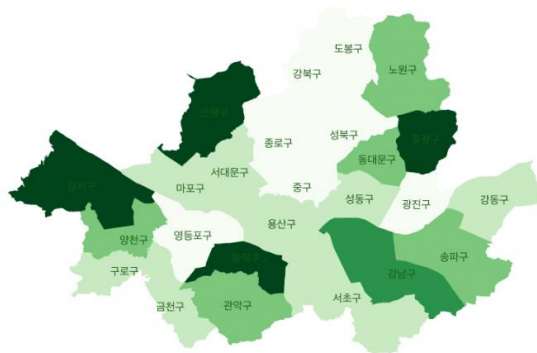
As above, it will be used to search the surrounding environment, and it is made for other criteria application.



4.2. Processing

Aim 1 : Count kindergarten for each district.

because I want to check the difference between each district in Seoul.

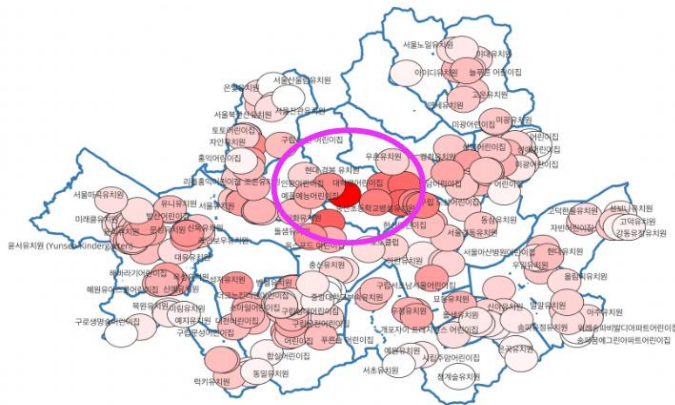


	ABC 자치구 ▼	123 k_total_cnt ↓ ▼
1	중랑구	24
2	동작구	24
3	은평구	20
4	강서구	20
5	강남구	17

TOP 5 districts

Aim 2 : Count surround facilities for each kindergarten(using radius)

It is a heat map derived from the number of nearby police station, post offices, marts, and parks as scores and tables created accordingly.



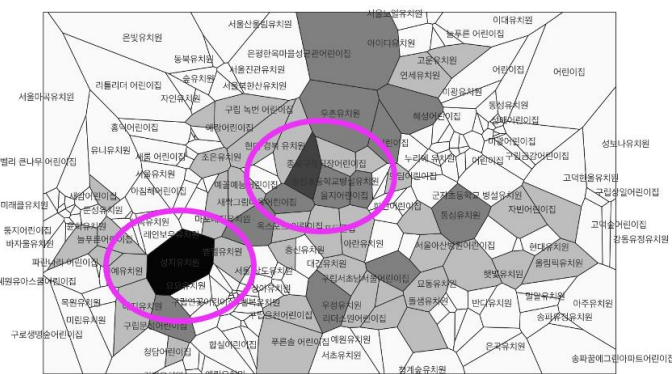
Most of high score kindergartens are located in '중구'

Aim 3: Count Surrounding facilities for each kindergarten(using Voronoi)

same as above. Consider the number of nearby police station, post offices, marts, and parks as scores.

123 ogc_fid	ABC name	123 frest_num	123 mart_num	123 police_num	123 post_num	123 total_score
169,750	성지유치원	6	6	8	17	37
163,384	종로구청청장어린이집	4	2	11	11	28
283,776	서계어린이집	5	1	8	7	21
62,125	어린이집	3	6	4	6	19
146,106	[NULL]	1	1	10	7	19
245,480	우촌유치원	2	3	6	8	19
478,897	동신유치원	3	1	5	10	19
431,007	돌샘유치원	4	2	3	8	17
486,086	[NULL]	2	0	5	10	17
48,798	반포퍼스티지하늘어린이집	3	3	5	5	16

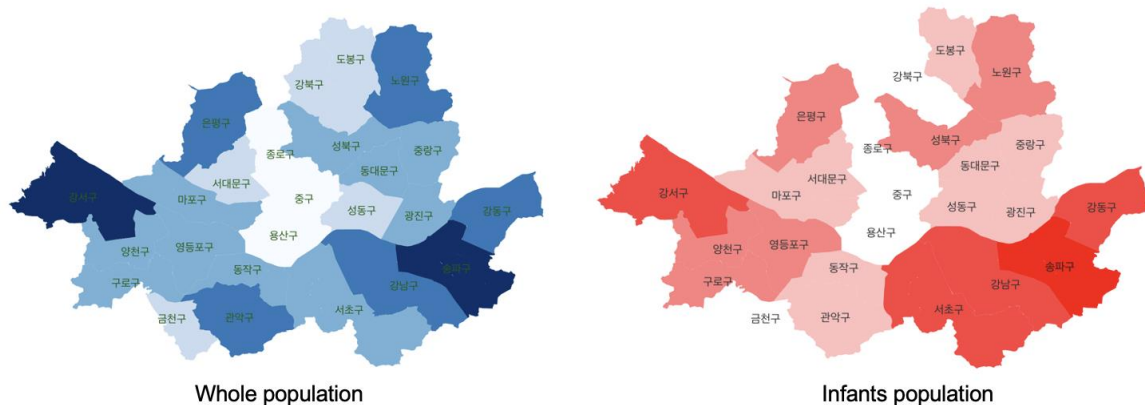
then I can draw a heat map below.



Most of high score kindergartens are located in '중구' and '영등포구'

Aim 4: draw the heat map of the population in Seoul by districts

you can see the population concentrated in the Seocho-gu and Ganam-gu line.



The darker the color, the larger the population.

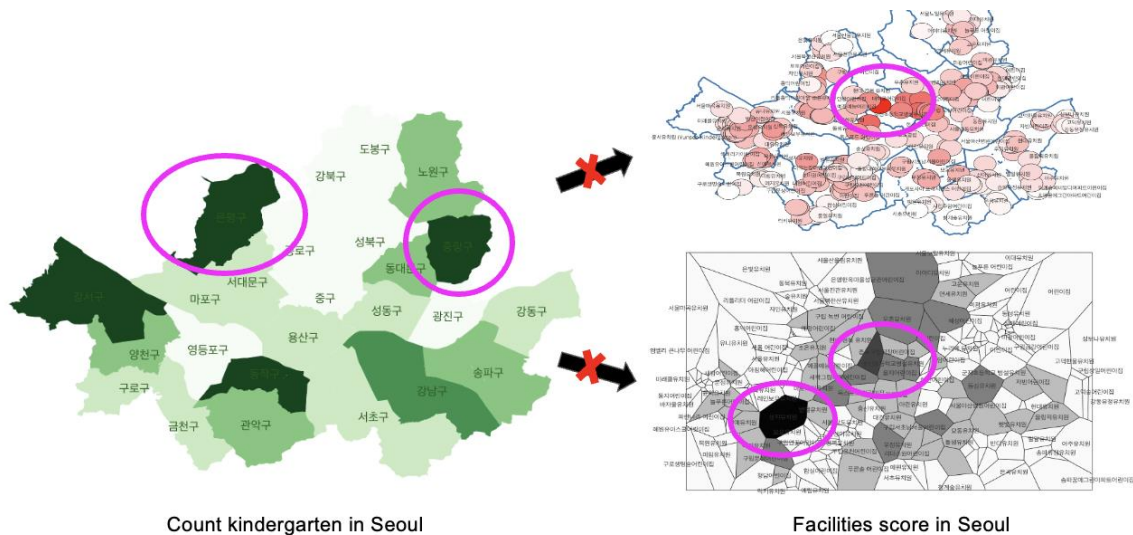
5. Analysis

Now I have all heat map from our tables.

let's summarize it into the analysis.

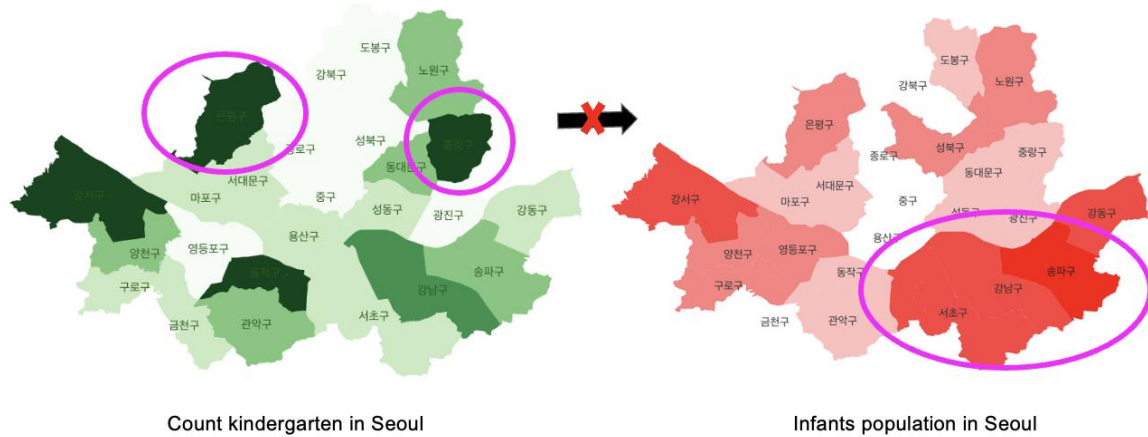
On the left is a heat map of the distribution of kindergartens in Seoul, and on the right is a heat map of the scores of surrounding facilities in Seoul.

Contrary to the expectation that there will be many kindergartens in places with many nearby facilities, I can see that kindergartens are concentrated in certain areas.

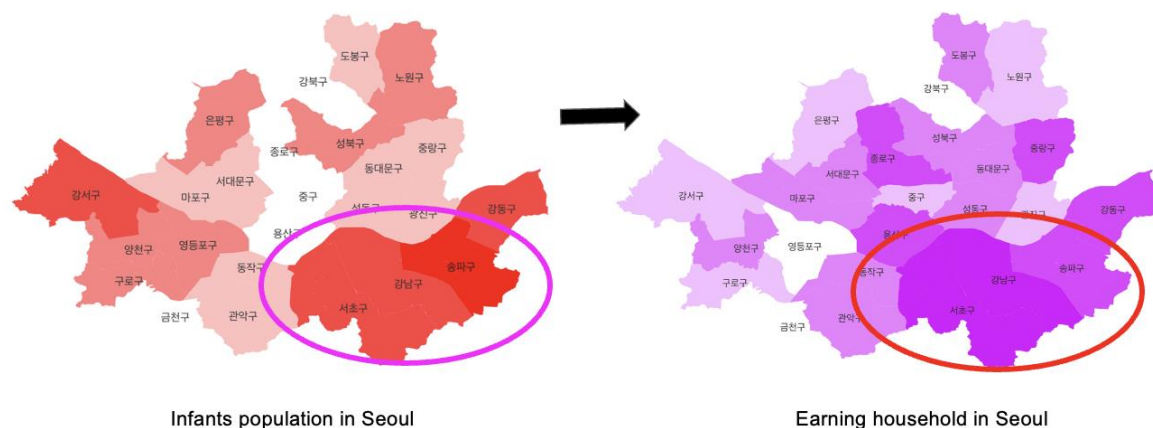


The following is a comparison with the distribution of infant population.

Contrary to the expectation that there will be a lot of infants in places with many kindergartens, it was concentrated in Seocho-gu and Gangnam-gu, even there are few kindergartens.



The expected high-income as a reason for the crowding in Seocho-gu was almost consistent with the distribution of the number of infants.



6. Conclusion

It was different from the actual data. Which was our first prediction, “The surrounding infrastructure greatly influence kindergarten choices.”

The problem was that parent wanted to raise their children in a high-income area

So, I can propose this kind of solution.

first of all, the number of kindergarten in areas where the competition rate is fierce such as Gangnam-Gu, Seocho-Gu, Songpa-Gu should be increased to resolve the current competition rate for kindergarten

admission.

A more fundamental solution is

1. to abandon the perception that children with parents who are currently prevalent high-income earners are in good relationships
2. promote balance development in all regions
3. create a perception that they can enter kindergartens with low competition rates.(I mentioned that if considering the overall level of Korea, the number of kindergartens is not insufficient.)

The above solution will lead to the expected effects below.

1. the resolution of excessive competition for kindergarten admission.
2. resolving the population concentration phenomenon caused by the competition for kindergarten admission.
3. balanced development through distribution of population.