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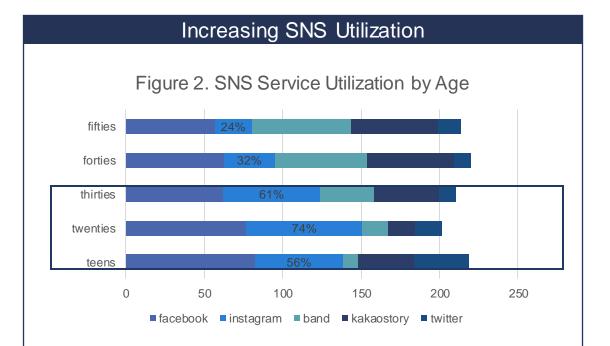
# 1. Introduction

Purpose & Outline

### Background



- "Gentrification", the phenomenon that indigenous people are pushed out due to the inflow of outsiders and money and the increase of rents as the underdeveloped area near the city is activated.
- Because of gentrification phenomenon, **Seoul marketing** areas' vacancy rate steadily increases, there areas' businesses are going downturn.
- Entrepreneurs worry about choosing a business location.



- Recently, people in their 20s and 30s actively share consumption tendencies **using SNS**, especially Instagram.
- Based on this, we will select a hot place and analyze the variety characteristics of the district to predict the district that will emerge as a hot place in the future.

### Purpose







# Establish countermeasures against gentrification problems

Providing location insight to future entrepreneurs in their 20s&30s

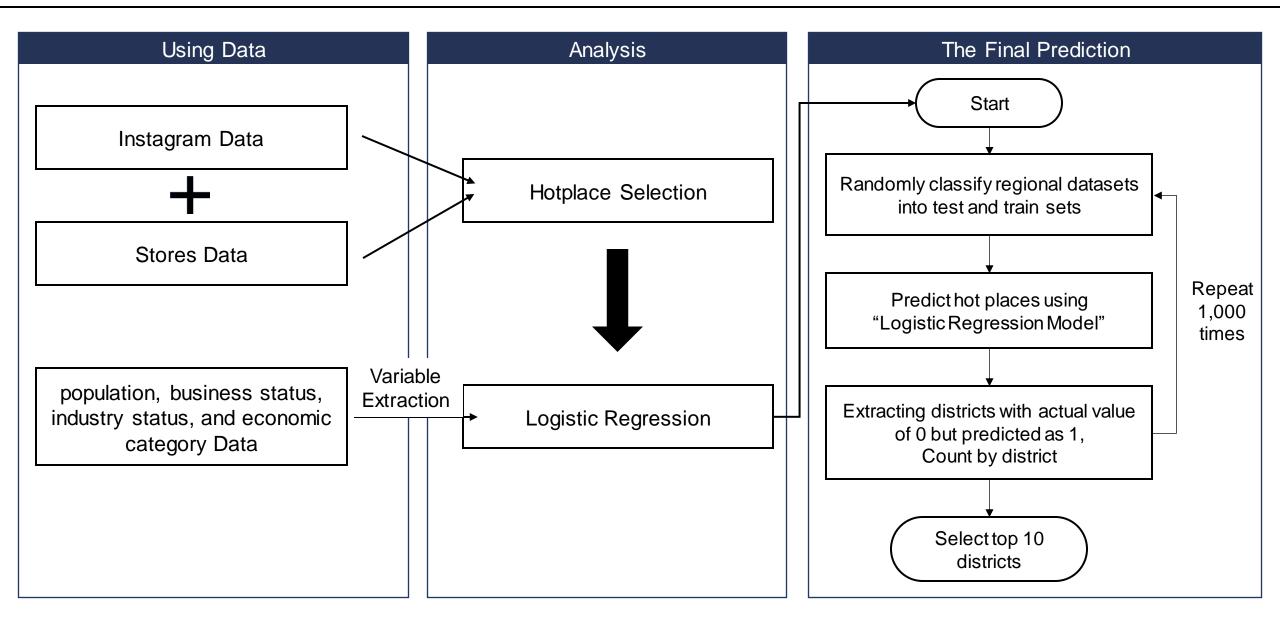
Help local revitalization

As a result of this study, it is expected that the prediction of the 'hot place' can be made in advance to establish the policy and the countermeasures against it.

The prediction of hot places would provide location insights to future entrepreneurs preparing to start a business.

By selecting highly correlated factors among the variables, provide insights to selected hot place candidates to help revitalize the region.

### Outline



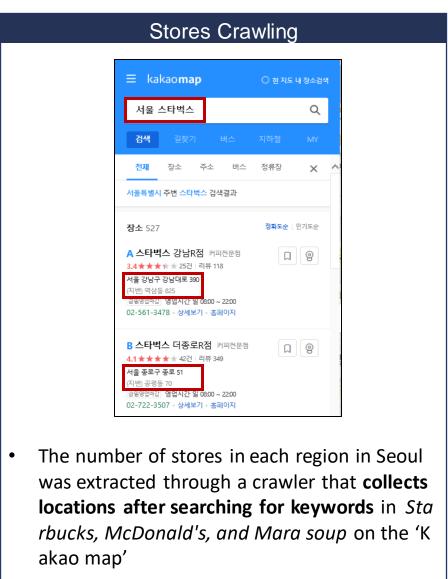
## II. Data Collection

Data for selecting Data for predicting

### Data for Selecting



- By searching for tags related to hot places (jmt, waiting, dating, restaurants, cages), we crawled location about 300~400 posts per tag
- the collected location was searched by the crawler on Kakao map, and only the location in Seoul was extracted and classified



### Data for Predicting

Collection for hot place analysis, prediction





Category	Data	
Population	Moving in, Moving out, single household, 2 households, 3 or more household, weekly population index, resident registration population	
Industry Status	Food, Living Service, Leisure, Sports, Academic Education, Accommodation	
Business	Operation Store / Closed Store	
Economy	GDP, Level Index	

Table 1. Extracted data by category

# III. Data Analysis

Hot Place Selection Variable Extraction Hot Place Prediction Algorithm

### Hot Place Selection

# District Scoring by Instagram Hash Tag

Figure 3. The number of each tag for each district

 The Instagram location hashtag is determined to be sufficient to select a representative hot place in Seoul through hash tag visualization.

### District Scoring by the Number of Stores

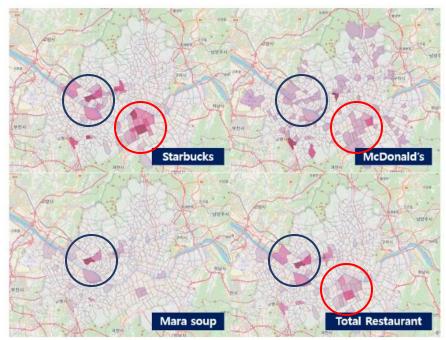


Figure 4. The number of each store for each district

- The number of Starbucks and McDonald's stores is similar to the Instagram hashtag, and is a good example of a representative hot place.
- In the case of Mara soup, Instagram hashtags appear in the **most crowded** places, and the importance of store weight should be weighted.

### Hot Place Selection

### District scoring

Because Total Mara soup stores are significantly smaller than Starbucks / McDonald's

Score of Store by region = Number of Marasoup stores per region \* 2+ Number of Starbucks per region + McDonald's per region



SNS post count by district = SNS score by district

Total score by distric =

SNS score by district \* 0.7 +store score by district \*0.3

Becuase it's about selecting districts that 2030 population "actually goes to"

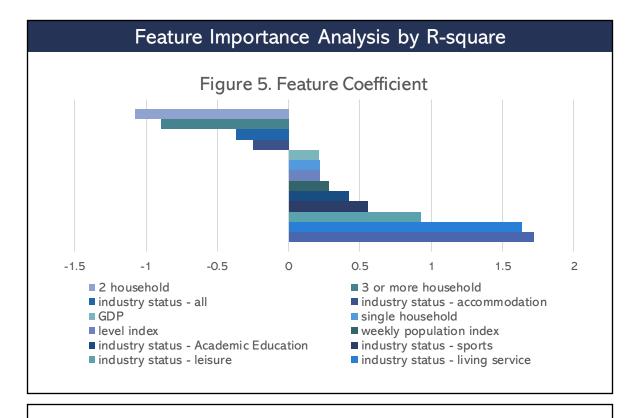
### Hot place selection result

rank	name	Total Score
1	서교동	37.9
2	신사동(강남)	31.7
3	역삼1동	17.8
4	종로1.2.3.4가동	16.7
5	여의동	13.6
6	신촌동	13.2
7	청담동	13.1
8	연남동	13.1
9	삼성1동	12.4
10	이태원1동	9.7
	•••	•••

rank	name	Total Score
275	상도4동	0
276	부암동	0
277	무악동	0
278	교남동	0
279	창신2동	0
280	창신3동	0
281	숭인2동	0
282	신당5동	0
283	황학동	0
284	다산동	0
	•••	•••

- The score difference between the top 40 districts and the bottom 384 districts was large.
- Out of a total of 424 districts, 40 districts(Seogyo-dong, Sinsa-dong, Yeoksam 1-dong, Jongro 1,2,3,4 Ga ...) are hot places.
- 384 districts(Burn 3dong, Suyu 3dong, ..)were classified as non-hot places.

### Variable Extraction



Exp 1. Variable Coefficient

$$z = D(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_d x_d$$

Exp 2. Logistic expression

$$p(y = 1|x) = g(z) = \frac{e^z}{1 + e^z} = \frac{e^{\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_d x_d}}{1 + e^{\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_d x_d}}$$

### Feature Selection

Dependent variable	Hot Place O/X			
Independent variable				
Population	1-person households, 2-person households, 3-person households or more, Weekly Population Index, Registration Population			
Industry Status	Food, Living Service, Leisure, Sports, Academic Education, Accommodation			
Business	Operation Store/ Closed Store			
Economy	GDP, Level Index			

Table 2. Final variables

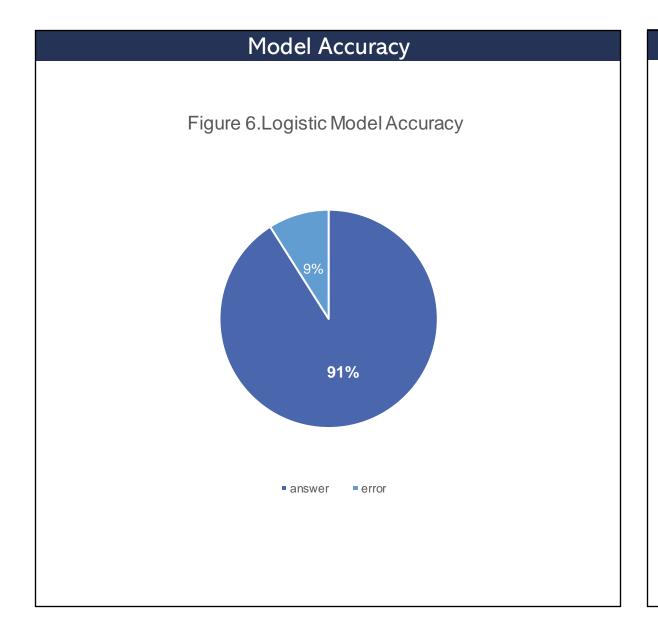
- Exp 1's  $\beta_0$ ,  $\beta_1$ ,  $\beta_2$ , ...,  $\beta_n$  value is the slope that the variable represents in logistic regression, which means the influence to dependent variable of each independent variable.
- Variables with  $\beta_n$  between 0.2 and 1.7 absolute values are selected as final independent variables.

### Hot Place Prediction Algorithm

### **Confusion Matrix Predict Non-Hot Place Predict Hot Place** Actually Non-Hot Place, **Actually Non-Hot Place But Predicted Hot Place Non-Hot Place** Hot Place, **Actually Hot Place Actually Hot Place** But Predicted Non-Hot Place Table 3. Confusion Matrix **β error**: Regions that Non-hot places but logistic model

- predicted to hot place
- The logistic model is suitable for binary classification since the above graph is derived by matching a kind of sigmoid function when calculating the probability.
- By training the logistic model 1,000 times, we can **extract** the regions predicted by the model as hot places from the non-hot places.

# IV. Result



### Predicted Hot Place

Figure 4. Top 10 districts

rank	name	count
1	중림동	340
2	삼청동	333
3	필동	308
4	남영동	294
5	종로 5,6가동	286
6	장충동	260
7	신당5동	255
8	대치4동	243
9	한강로동	222
10	광희동	213

# V. Conclusion

### Pil-dong

### 서울시, 도심 실핏줄 골목길 재생사업 본격 추진

조성신 기자 | 입력 : 2018.02.08 10:24:35 🔘 0



# "Alleyway revitalization with 'Seoul-type alleyway recycling business' "

서울시는 도시의 역사와 함께해온 역사문화유산이자 삶의 공간인 골목길을 일·삶·놀이가 어우러진 곳으로 재생하는 '서울현 골목길 재생사업' 응 올해부터 보격 초지하다고 8일 밝혔다.

`서울형 골목길 재생사업'은 도시재생활성화지역 등 일정 구역을 정해서 '면' 단위로 재생하는 기존 도시재 생사업과 달리 골목길을 따라 1km 이내의 현장 일착형 소규모 방식의 '선' 단위 재생 개념을 도압해 추진한다 당장 낮추기, 골목 마당 공유, 내 집 수선하기 같은 사업도 주민 주도로 함께 진행한다.

시는 지난해 8월에 착수한 `서울형 골목길 재생 기본계획` 용역을 3월경에 마무리하고 오는 5월까지는 골목 김 재생사업 가이드라인을 마련해 기본계획을 수립한 계획이다.

6월에는 자치구 공모를 통해 사업대상지를 추가 선정하고 재생사업을 본격 확대 추진한다. 기본계획 수립과 병행해 시는 용산구와 성복구 2곳 골목길을 시범사업지로 선정하고 5월 중 실행계획을 수립, 연내 사업에 착수할

### Junglim-dong

### 전통과 핫플레이스가 공존하는 곳…'반전 매력' 중림동 뜬다

988 2020 01 to 18:05 | 486 2020 01 11 15:10 | XRY A

오랫동안 과거에 머물러 있던 중림동이 크게 달라지고 있다. 2017년 봄 서울역 고가가 '서울로7017로 탈바꿈한 이후 해마다 동네의 모습이 바뀐다. '남대문시장과 회현동에서 시작되는 고가 보행로가 사람들의 발길을 중림동과 만리동으로 쉽게 이어준 덕분'이라고 상인들은 말한다.

"Young people have been crowded since the opening of Seoulro 2017 and 'hot places' including restaurants appears in every alley"

동네로 유입되는 사람들이 늘면서 맛집 등 '핫플레이스'들이 골목마다 등장했다. 독특한 콘셉트의 식당과 카페엔 주중에도 젊은이들이 붐빈다. 양식당 서울부띠끄, 퓨전레스토랑 동네로 유입되는 사람들이 늘면서 맛집 등 '핫플레이스'들이 골목마다 등장했다. 독특한 콘셉트의 식당과 카페엔 주중에도 젊은이들이 붐빈다. 양식당 서울부띠끄, 퓨전레스토랑 베리스트릿키친 등이 대표적이다. 설렁탕집 중림장, 닭꼬치 가게인 호수집 등 노포(老鋪)들은 전국적인 맛집으로 떠올랐다. 근대 서울의 모습과 젊은이들이 즐겨 찾는 '핫플레이스'가 공존하는 곳, 중림동이 서울 도심 명소의 하나로 거듭나고 있다.



남영동골목여행에서만난모듬스테이크,숙대입구맛집'서지스테이크'

용 최태인기자 I O 승인 2018.12.13 15:45 I O 댓글 0

"With the addition of restaurants and cafes, it is at tracting new attention as a hot place."

서울에서 만날 수 있는 대표적인 골목길로는 용산구 이태원동의 경리단길, 마포구 망원동의 망리단길, 관악 구 서울대 입구의 사로수길 등이 있다. 용산구 남영동의 열정도도 새롭게 주목받고 있는 핫플레이스다.

남영동 열정도는 숙대입구역과 남영역 인근 지역을 일컫는다. 이 지역이 눈길을 끌기 시작한 것은 청년사업가 들이 지역 상권 살리기에 발벗고 나서면서다.

중소규모의 인쇄업체가 모여 있던 골목길에 감각과 위트가 가득한 식당, 카페 등이 하나둘 들어서면서 사람들 의 발걸음을 모으고 있다.

Recent developments in the region and the formation of commercial zones are likely to be hot places.

# Thank you.