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주 제 소 개

나만의 여행지를 추천해주는 *유럽 여행 도우미* 













wordcloud

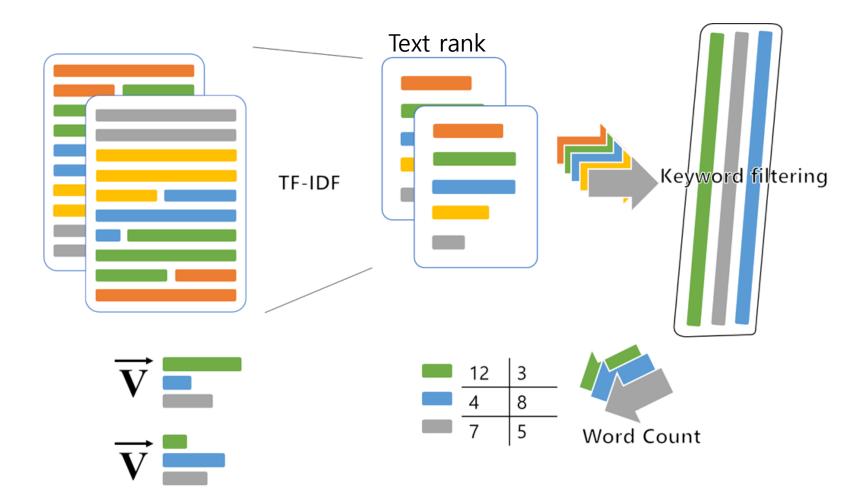
Insight



# 데이터수집 W | K |

- Wikipedia에서 스페인, 포르투갈 관광지 문서 수집
- 문서들을 바탕으로 관광지들간의 관계를 파악해 벡터화 (loc2vec)
- 추가적으로 위치와 거리 정보를 활용해 다음 관광지 추천

	place	nature	city	star	review	short_info	long_info	f_wiki	s_wiki
0	Buen Retiro Park	스페인	Madrid	4.6	24568	Vast 19th-century park with fountains	Expansive, 19th- century park with boating lake	The Buen Retiro Park (Spanish: Parque del Buen	The Buen Retiro Park (Spanish: Parque del Buen
1	Royal Palace of Madrid	스페인	Madrid	4.4	6566	Regal 2,000-room palace, armory & garden	18th-century, ridge- top palace for state occas	The Royal Palace of Madrid (Spanish: Palacio R	The Royal Palace of Madrid (Spanish: Palacio R
2	Plaza Mayor, Madrid	스페인	Madrid	4.3	17375	Madrid's vibrant main square	Cafes & restaurants line the arches of thi	The Plaza Mayor (English Main Square) was buil	The Plaza Mayor (English Main Square) was buil
3	Museo Nacional Del Prado	스페인	Madrid	4.6	13871	World-class European art collection	Art museum with Velazquez, Goya and El Greco m	The Prado Museum (Spanish pronunciation: [mu? s	The Prado Museum (Spanish pronunciation: [mu? s
4	Puerta del Sol	스페인	Madrid	4.3	26932	Vast pedestrianized public square	Public square with an equine statue of King Ca	The Puerta del Sol (Spanish for "Gate of the S	The Puerta del Sol (Spanish for "Gate of the S



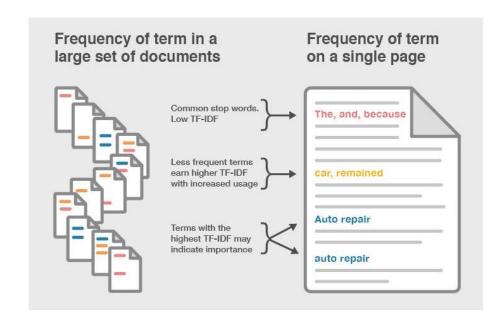
# 1 **TF-IDF**

문서 내에서 단어의 중요도를 파악하자

- 중요도가 높은 단어는 문장 내에서 자주 등장한다
- 한 단어가 문서 전체에 계속 등장하면 중요도가 떨어진다

→모든 문서에 흔하게 나타나는 단어는 제외시킴

 $W_{i,j}=tf_{i,j}\times log(N/df_i)$ 



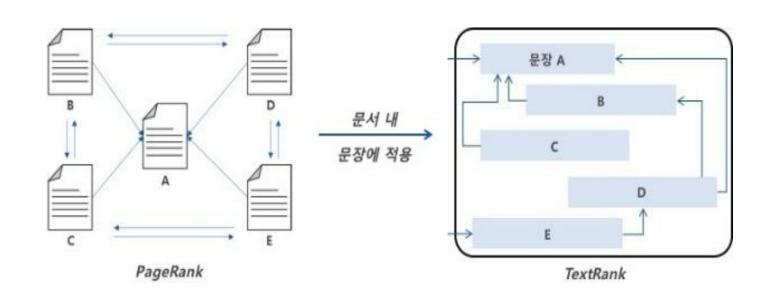
# 2 **Text Rank**

문서 내에서 키워드들을 뽑아내보자

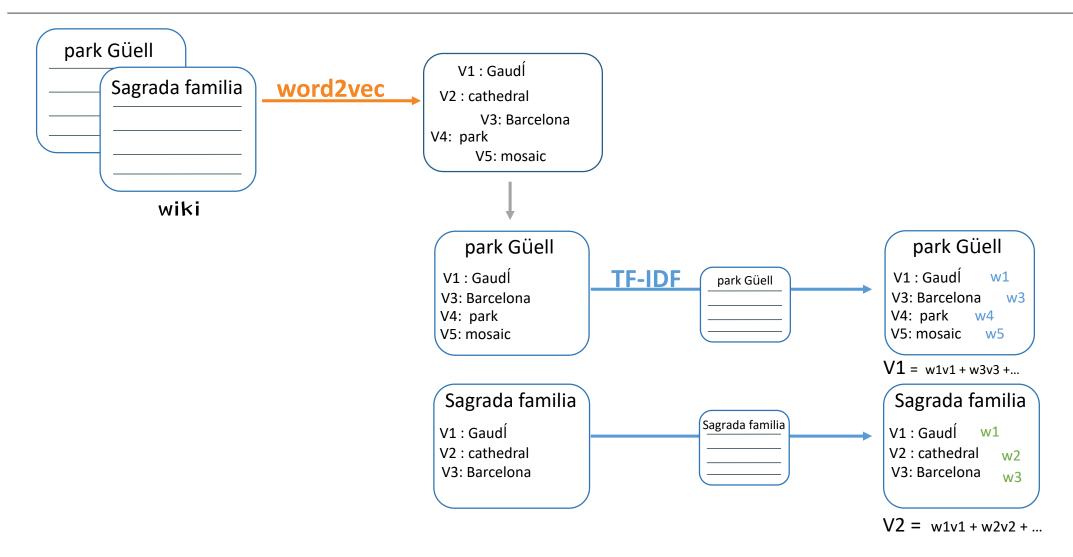
- 중요도가 높은 단어는 여러 문장과 link되어 있다

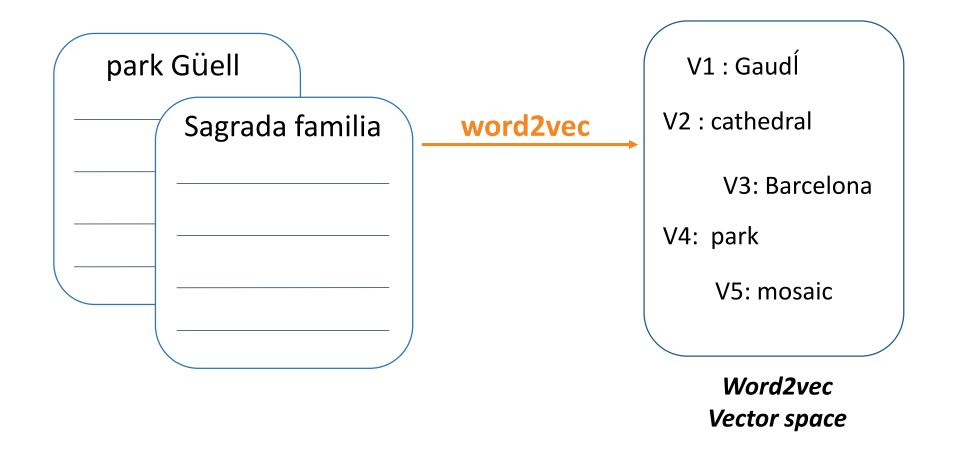
$$TR(V_i) = (1-d) + d * \sum_{V_j \in In(V_i)} rac{w_{ji}}{\sum_{V_k \in Out(V_j)} w_{jk}} TR(V_j)$$

Wji: 단어 i와 j 사이의 가중치 ←TF-IDF



## $w \circ r k f l \circ w$ 2 nd trial





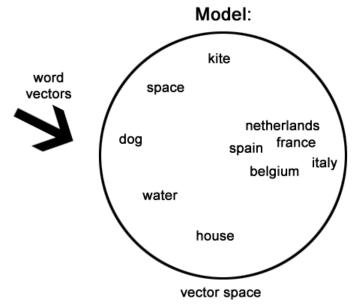
# 1 word2vec

단어가 가지는 의미 자체를 벡터로 나타내자

- 같은 문장에 등장하는 단어는 의미가 비슷하다
- man woman + king = queen

# Input: one document

Lorem ipsum dolor sit amet, consetetur sadipscing elltr, sed diam nonumy eirmod tempor invidunt ut labore et dolore magna aliquyam erat, sed diam voluptua. At vero eos et

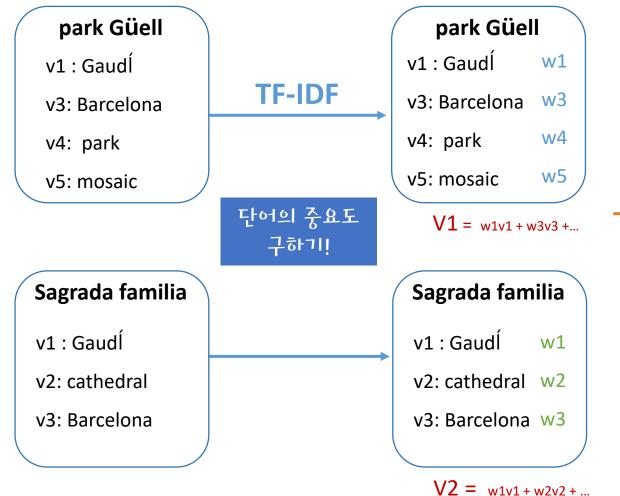


#### most\_similar('france'):

spain	0.678515
belgium	0.665923
netherlands	0.652428
italy	0.633130

highest cosine distance values in vector space of the nearest words

## $w \circ r k f l \circ w$ 2 nd trial

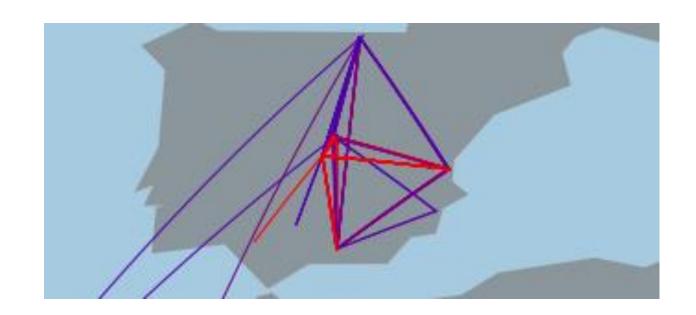


# 지도 상 거리로 필터링



LOC2vec

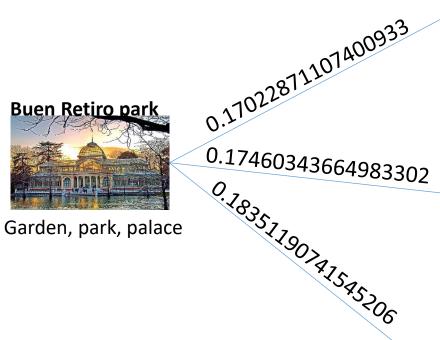
Loc2vec 벡터 공간 안에서 거리가 가까울 수록 유사한 관광지



# Euclidian distance

# $P_{1}(x_{p}, y_{1})$ Euclidean distance (d) = $\sqrt{(x_{2}-x_{1})^{2} + (y_{2}-y_{1})^{2}}$

### Euclidian distance를 이용해 query에서 가장 가까운 관광지 5개를 제시





Madrid, gardens, park



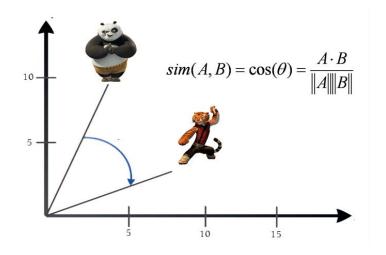
Reasons, line, location

#### Plaza de Bib-Rambla



Building, line, city

# Cosine similarity



Cosine similarity를 이용해 query에서 가장 가까운 관광지 5개를 제시

0.46313844006755767

Parque de el Capricho

Madrid, gardens, park



Garden, park, palace

Parque de Atraciones de Madrid 0.38834654597694782 0.31460515328705>



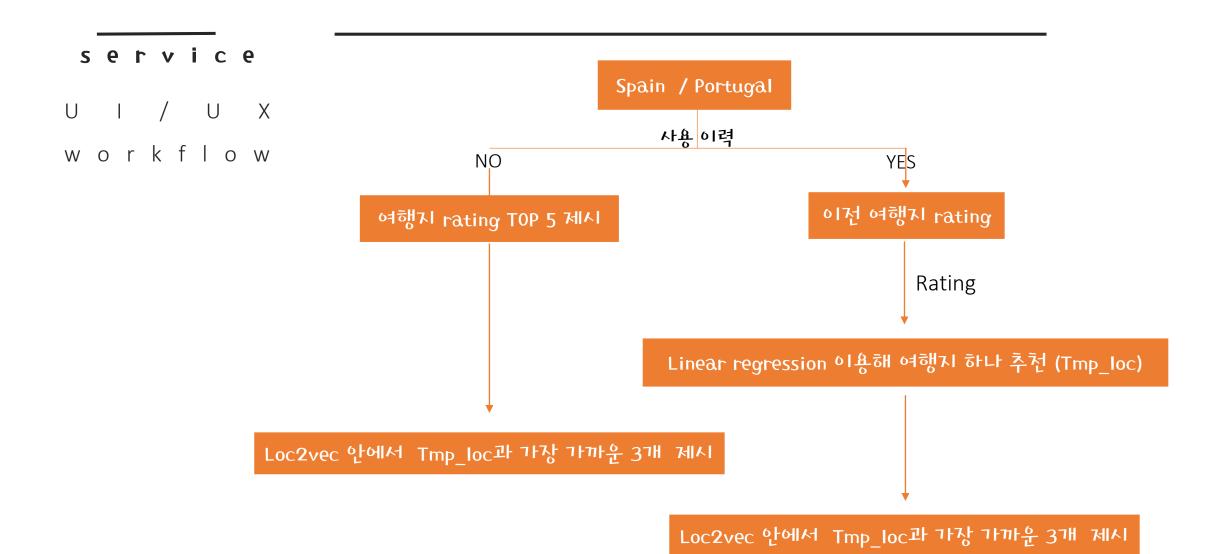
Park, attractions, zone

Plaza de Espana



Plaza, espana, parque

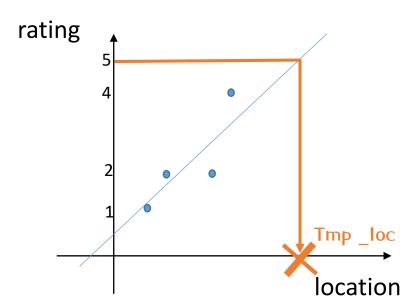




1 Linear Regression

$$X1$$
  $X2$   $X3$  ...  $X100$  Rating A (a1 a2 a3 ... a100) r1 B (b1 b2 b3 ... b100) r2 C (c1 c2 c3 ... c100) r3 D (d1 d2 d3 ... d100) r4 회가  $\beta$ 1  $\beta$ 2 ...  $\beta$ 100

$$5 = \beta 1 \times 1 + \beta 2 \times 2 + \dots + \beta 100 \times 100$$



#### service

U I / U X workflow

#### **Royal palace of Madrid**



Palace, royal, charles

#### **Prada Museo**



Historician, nationalisation

#### Plaza Mayor, Madrid



Plaza, Philip, square

#### **Buen Retiro park**



Garden, park, palace

Puerta del sol



Square, city, plaza

#### **Madrid Rio**



Reasons, line, location

#### Gran via



Street, avenue

