

Retirement Recommender

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Problem



- ❑ Retirees have so many options to choose from when it comes to where to retire.
 - ❑ There are so many cities in the US
 - ❑ The dataset used in this project contains more than 300 US cities.
 - ❑ Retirees want an easy and painless transition to retirement
- ❑ Moving to a new city is hard
 - ❑ Having amenities and attractions similar to your previous city is useful
 - ❑ Choosing a new place is a hassle

Question



- Can we make it easy for those people who are choosing a new place to live by allowing them to enter a city of their choice, and provide multiple suggestion of similar cities that might also be a good idea for retirement?
- Can we reduce the number of options they have?

Dataset

- US top cities by population data from Wikipedia 314 cities
- From this data, we used three data sources
 - Population data from Wikipedia
 - Geographical Latitude and Longitude data from geocoders
 - Amenity and attraction data from Foursquare API
- 517 features
 - 516 amenity and attraction categories



Data transformation

- ❑ Start with top 314 cities in US
- ❑ Add population data
- ❑ Add geographic data
- ❑ Create new data frame with venues for each city listed
- ❑ Convert the venue category into a numerical value
- ❑ Get the mean for each category for each city
- ❑ Make the location the index of the table



Data Analysis and Modeling



KMeans Clustering algorithm

K= 40 clusters centers

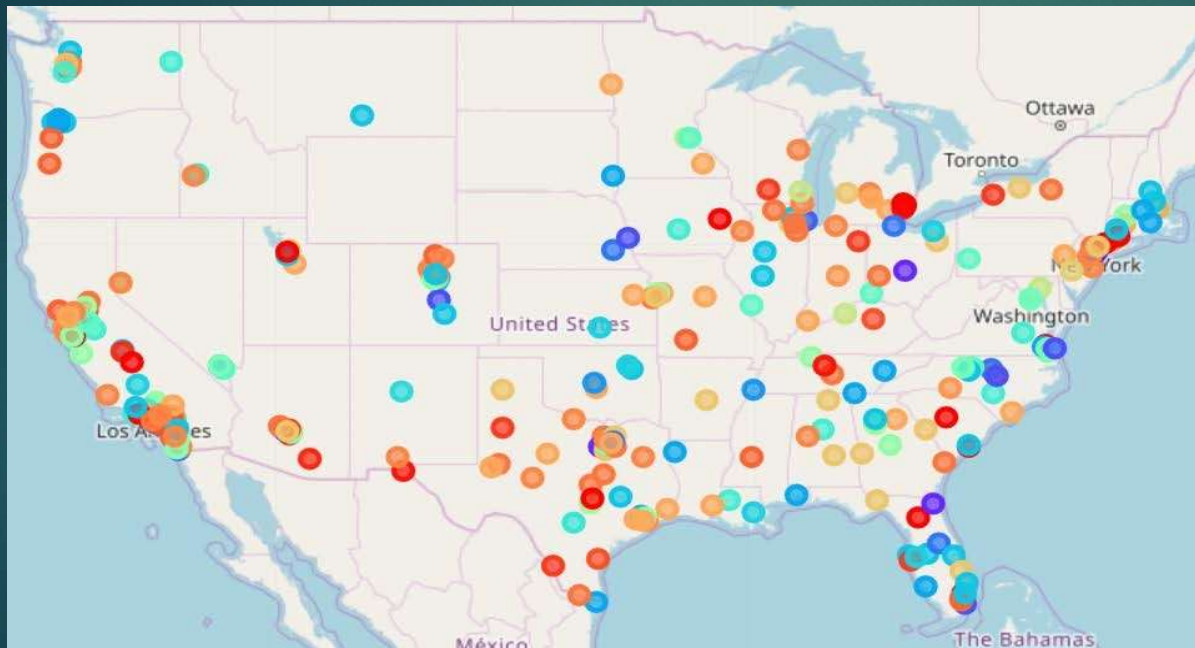
N = 312 cities

Features = 517 (population data +
Foursquare data)

Output clusters were filtered
to only contain the groups
that have 5+ cities



Results



N = 266 cities labeled

Filtered by cluster label
size > 5 and available
Foursquare data

Results (Cont)

Cluster Labels			
0	19	19	19
5	6	6	6
7	9	9	9
10	14	14	14
12	36	36	36
13	8	8	8
16	13	13	13
18	7	7	7
20	11	11	11
22	15	15	15
29	20	20	20
31	28	28	28
33	32	32	32
34	25	25	25
35	14	14	14
37	9	9	9

40 clusters total

Filtered where
cluster size > 5

Providing enough
recommendations

Cluster Labels		Location
0	31	Abilene, TX
1	29	Akron, OH
2	14	Albuquerque, NM
3	22	Alexandria, VA
4	33	Allen, TX
5	31	Allentown, PA
6	29	Amarillo, TX
7	27	Anaheim, CA
8	7	Anchorage, AK
9	31	Ann Arbor, MI
10	12	Antioch, CA
11	13	Arlington, TX
12	31	Arvada, CO
13	31	Athens, GA
14	23	Atlanta, GA

Example
cluster
labels

Output

```
cities_final2.loc[cities_final2['Location'] == 'Miami,FL']
```

	Cluster Labels	Location	City Latitude	City Longitude
15064	5	Miami,FL	25.774266	-80.193659

```
cities_final2.loc[cities_final2['Cluster Labels'] ==5]
```

	Cluster Labels	Location	City Latitude	City Longitude
4894	5	Colorado Springs,CO	38.833958	-104.825349
13663	5	Long Beach,CA	33.769016	-118.191605
15064	5	Miami,FL	25.774266	-80.193659
17519	5	Omaha,NE	41.258746	-95.938376
19791	5	Raleigh,NC	35.780398	-78.639099
26167	5	Virginia Beach,VA	36.852984	-75.977418

Input: city in dataset
Ex: Miami, FL

Output: similar cities in
the US with the same
cluster label