

# 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

N = 312

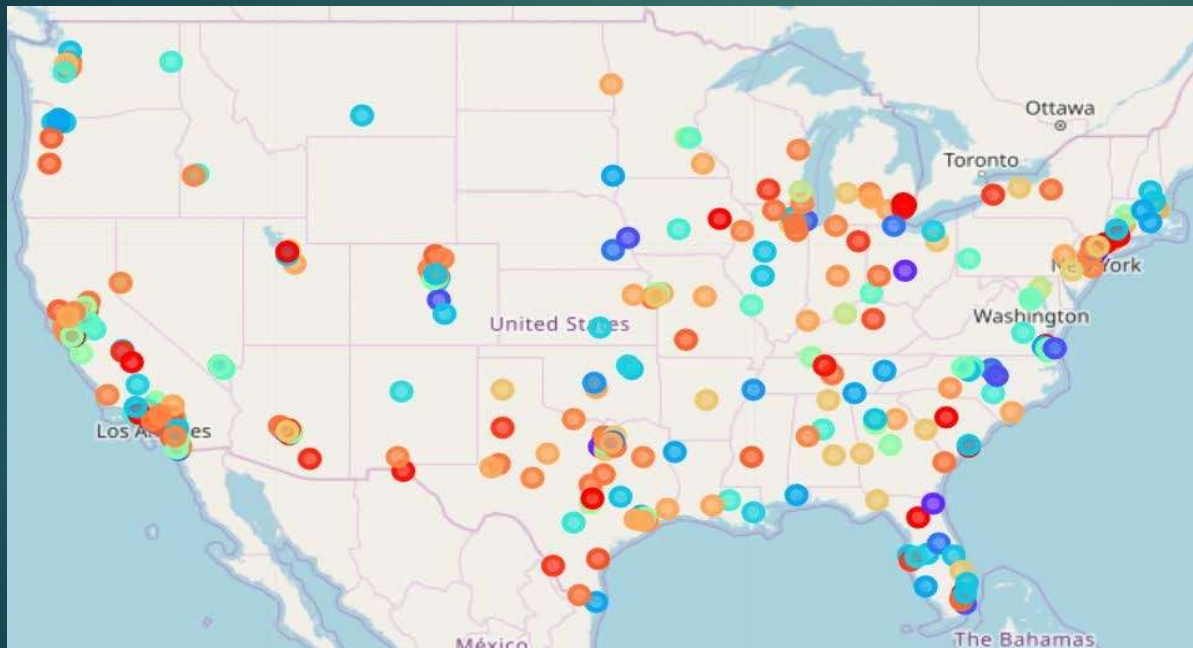
Features = 517

Output groups 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

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