

I. Introduction

Business Problem





Target Audience





 $\label{lem:http://www.gramercyparkblockassociation.org/images/neighborhood_news_559_6_1o_14.html \\ https://www.jacksonvilleu.com/blog/business/business-analytics-market-research-analyst-job-description-salary/$

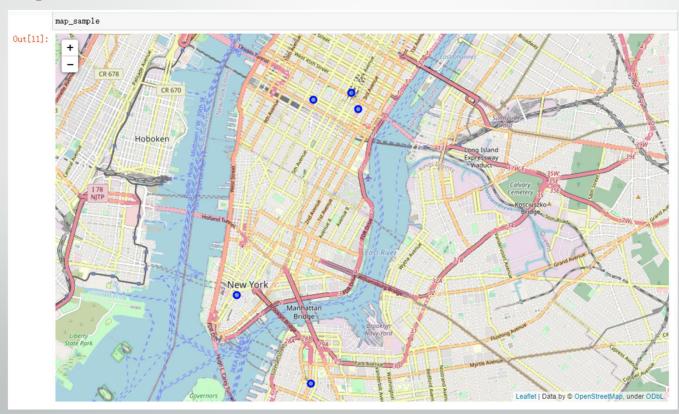
II. Data Description

Determine the Research Subjects

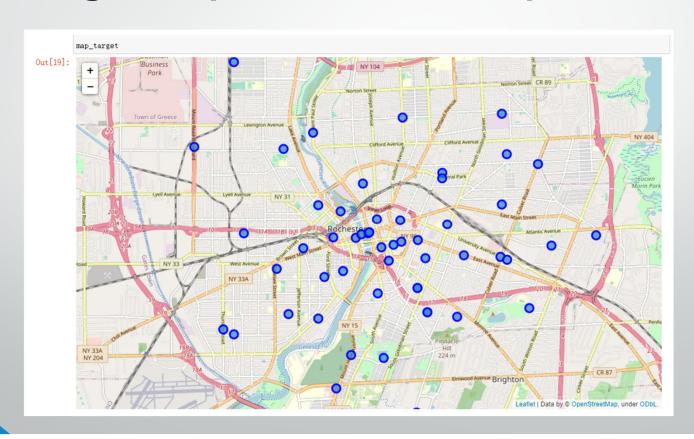


http://www.sactownmag.com/Whats-Cooking/2019/Shake-Shack-Sacramento-Ice-Blocks/

Sample Stores Proven to be Successful



Target City for Business Expansion



Venue Data

Out[8]:												
	0	Name	Center Latitude	L	Center ongitude	Venue	Venue Latitude	Venue Longitude				
	0	Shake Shack in NYC	40.71056	-74	4.009014	Shake Shack	40.710703			ger Joint		
	1	Shake Shack in NYC	40.71056	-74	4.009014	Anthropologie	40.710618			Vomen's Store		
	2	Shake Shac (in NY	Out[20]:		Name	Center Latitude	Center Longitude		Venue	Venue Latitude	Venue Longitude	Venue Category
	3	Shake Shac in NY Shake Shac in NY	-	0	Convention District	43.156389	-77.609167		n's The	43.156566	-77.608455	Steakhouse
	4			1	Convention District	43.156389	-77.609167	Roc Riverside	chester e Hotel	43.156826	-77.609954	Hotel
				2	Convention District	43.156389	-77.609167	Sta	rbucks	43.156616	-77.608549	Coffee Shop
				3	Convention District	43.156389	-77.609167	Hyatt Re	egency	43.156469	-77.608561	Hotel
				4	Convention District	43.156389	-77.609167		Focus ounge	43.156549	-77.608646	Bar

Data Cleaning and Preprocessing

	N	lame	Airport Lounge	BB Joir	н	akery	Bar	Basketball Stadium	Bubble Tea Shop	a Burge		fé Clothi Sto	
0	S	hake hack in NYC	0.013889	0.01388	39 0.0	13889 0.	.027778	0.013889	0.01388	9 0.13888	9 0.01388	9 0.0138	
			Name	Airport Lounge	BBQ Joint	Baker	y Bar	Basketball Stadium	Bubble Tea Shop	Burger Joint	Café	Clothing Store	t
	0		C Streets aborhood	0	0	0.00000	0.2	0	0	0.000000	0.000000	0	
	1	Bu	ill's Head	0	0	0.00000	0.0	0	0	0.000000	0.000000	0	
	2		anging of Scenes	0	0	0.00000	0.0	0	0	0.000000	0.000000	0	
	3	Colle	ege Town	0	0	0.076923	3 0.0	0	0	0.076923	0.076923	0	
	4	Co	onvention District	0	0	0.00000	0.2	0	0	0.000000	0.000000	0	←

III. Methodology

Identify the Optimal K Value

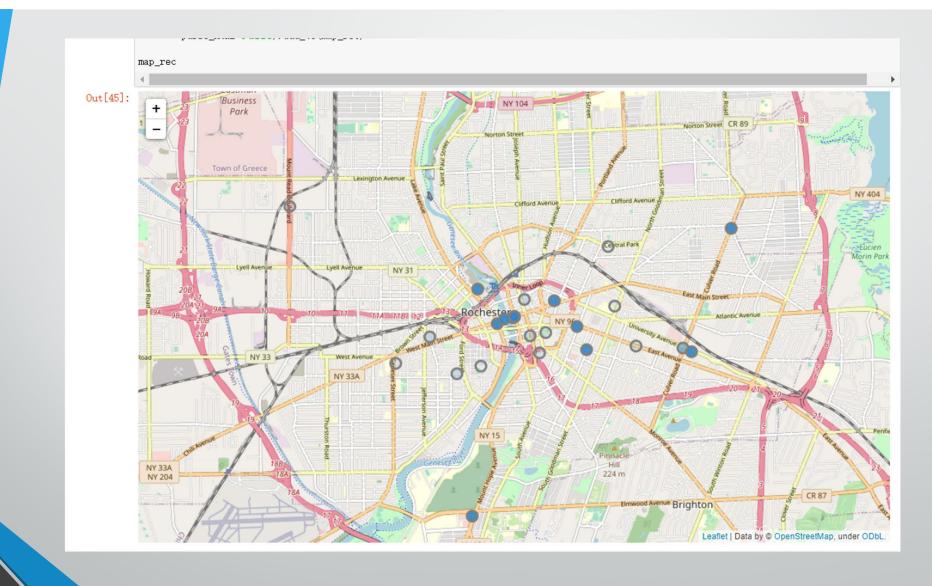
```
In [31]: kmax = 20
           errors = []
           for k in range(1, kmax+1):
              error = 0
              kmeans = KMeans(n_clusters=k, random_state=0).fit(k_clustering)
              centroids = kmeans.cluster_centers_
              label = kmeans, labels
              for i in range(0, len(k_clustering)):
                  groupid = label[i]
                  centroid = centroids[groupid]
                  dist = linalg.norm(k_clustering.iloc[i].to_numpy() - centroid)
                  error += dist
              errors.append(error/len(k_clustering))
           print (errors)
In [32]: x_axis = range(1, kmax+1)
          plt.plot(x_axis, errors)
          plt. show()
           0.5 -
           0.3
           0.2
                  2.5 5.0 7.5 10.0 12.5 15.0 17.5 20.0
```

```
In [33]: from sklearn.metrics import silhouette_score
          sil = []
          kmax = 20
          # dissimilarity would not be defined for a single cluster, thus, minimum number of cluster
          for k in range(2, kmax+1):
            kmeans = KMeans(n_clusters=k, random_state=0).fit(k_clustering)
            labels = kmeans, labels
            sil.append(silhouette_score(k_clustering, labels, metric = 'euclidean'))
In [34]: x_axis = range(2, kmax+1)
          plt.plot(x_axis, sil)
          plt.show()
           0.30
           0.28
           0.26
           0.24
           0.22
           0.20
                             7.5 10.0 12.5 15.0 17.5 20.0
```

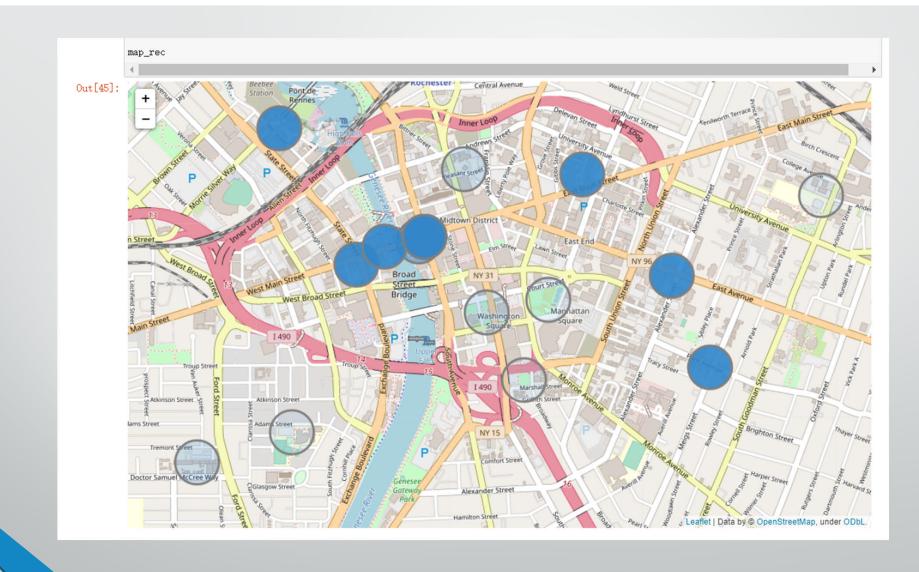
K-means Clustering Clifford Avenue NY 31 NY 33A 224 m Elmwood Avenue Brighton

Recommendation

IV. Results



V. Discussion



VI. Conclusion

