Capstone Project - The Battle of Neighborhoods

Part 1: Introduction

San Francisco is one of the financial, cultural and technology centers on the west coast, it has population close to 900,000. San Francisco is a diverse and culturally rich city, where you expect people to live in different lifestyles across different neighborhoods. Indeed, some of its neighborhoods are cozy and relaxing, while some others are busy and commercialized. Therefore, it is very important for either a business to choose where to open a new store or a person to pick where to live with his or her lifestyle.

I want to utilize Foursquare location to identify venues within each neighborhood, and then use venues' frequencies within each neighborhood to create clusters that provide insightful information for business and people to choose the target neighborhoods to open a new business or live with a desired lifestyle.

Part 2: Data Acquisition and Cleaning

1. Data Sources

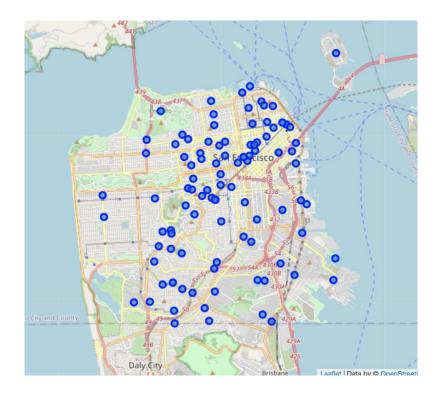
The data is from the following sources:

- a. San Francisco neighborhood list: Wikipedia SF Neighborhoods;
- b. Location data: Opencage Geocoder;
- c. Venues data: Foursquare;
- 2. Data Cleaning
 - a. For San Francisco neighborhood data, I used "mw-headline" class in Beautiful Soup Python library to extract the neighborhood list from the Wikipedia website;

	Neighborhood
0	Alamo Square
1	Anza Vista
2	Ashbury Heights
3	Balboa Park
4	Balboa Terrace
114	West Portal
115	Western Addition
116	Westwood Highlands
117	Westwood Park
118	Yerba Buena

119 rows × 1 columns

df



b. For the location data, I used the Opencage Geocoder Python library and free API key from Opencage to create a data frame that includes the latitudes and longitudes for each of the neighborhood in San Francisco. I have noticed the data frame generated from the Opencage Python library are not 100% accurate, so I deleted the rows that contained inaccurate location data;

	Neighborhood	lat	Ing
0	Alamo Square	37.776360	-122.434688
1	Anza Vista	37.780836	-122.443149
2	Ashbury Heights	37.775599	-122.448068
3	Balboa Park	37.721427	-122.447547
4	Bayview	37.728889	-122.392500
101	West Portal	37.741141	-122.465634
102	Western Addition	37.779559	-122.429810
103	Westwood Highlands	37.725726	-122.458199
104	Westwood Park	37.725726	-122.458199
105	Butchertown	37.784827	-122.727802

106 rows × 3 columns

df

c. Once I had the location data, I used <u>Foursquare</u> to generate 200 venues with a radius of 1,000 meters from the coordinate for each neighborhood. Then I created a

funcation called getNearbyVenues, which generates a data frame that contains each neighborhood and venue information.

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category
0	Alamo Square	37.776360	-122.434688	Alamo Square	37.776045	-122.434363	Park
1	Alamo Square	37.776360	-122.434688	Alamo Square Dog Park	37.775878	-122.435740	Dog Run
2	Alamo Square	37.776360	-122.434688	Painted Ladies	37.776120	-122.433389	Historic Site
3	Alamo Square	37.776360	-122.434688	The Independent	37.775573	-122.437835	Rock Club
4	Alamo Square	37.776360	-122.434688	The Mill	37.776425	-122.437970	Bakery
5102	Westwood Park	37.725726	-122.458199	Old Balboa Reservoir Berm	37.727362	-122.453289	Dog Run
5103	Westwood Park	37.725726	-122.458199	Orchids Cafe	37.723141	-122.453710	Cha Chaan Teng
5104	Westwood Park	37.725726	-122.458199	Wiley's No Limit Liquor & Food Mart	37.723354	-122.453505	Liquor Store
5105	Butchertown	37.784827	-122.727802	JazzDeck	37.784800	-122.727800	Music Venue
5106	Butchertown	37.784827	-122.727802	Newly Formed Emerald Park to Benefit San Franc	37.783734	-122.729690	Park

5107 rows × 7 columns

sf_venues

3. Data Exploration

After I obtained venue data from each neighborhood, I wanted to count the number of venues within each neighborhood and to display top 5 neighborhoods, and I found the following 5 neighborhoods have the largest numbers of the venues:

	Neighborhood	count
89	South of Market	196
37	Hayes Valley	100
35	Golden Gate Heights	100
13	Chinatown	100
9	Butchertown (Old and New)	100

I created dummy variables for the venue categories and calculated the mean for each venue category within each neighborhood to reveal how frequent those venue categories appeared.

Neighborhood	Yoga Studio	ATM	Acai House	Accessories Store	Acupuncturist	Adult Boutique	Alternative Healer	American Restaurant	Animal Shelter		Video Store	Vietnamese Restaurant	Vineyard	Wareho
Alamo Square	0.013514	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.000000	0.0	
Anza Vista	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.000000	0.0	
Ashbury Heights	0.038462	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.038462	0.000000	0.0	
Balboa Park	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.055556	0.0	
Bayview	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.000000	0.0	
Vista del Mar	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.010526	0.0		0.000000	0.000000	0.0	
West Portal	0.023810	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.000000	0.0	
Western Addition	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.000000	0.0	
Westwood Highlands	0.056604	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.037736	0.0	
Westwood Park	0.056604	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	0.0		0.000000	0.037736	0.0	
	Alamo Square Anza Vista Ashbury Heights Balboa Park Bayview Vista del Mar West Portal Western Addition Westwood Highlands Westwood	Alamo Square	Alamo Square	Alamo Square O.013514 O.0 O.0 Anza Vista 0.000000 0.0 0.0 Ashbury Heights 0.038462 0.0 0.0 Balboa Park 0.000000 0.0 0.0 Bayview 0.000000 0.0 0.0 Wista del Mar 0.000000 0.0 0.0 West Portal 0.023810 0.0 0.0 Western Addition 0.000000 0.0 0.0 Westwood Highlands 0.056604 0.0 0.0 Westwood 0.056604 0.0 0.0	Alamo Square O.013514 O.0 O.0 O.0 Alamo Square 0.013514 0.0 0.0 0.0 Anza Vista 0.000000 0.0 0.0 0.0 Ashbury Heights 0.038462 0.0 0.0 0.0 Balboa Park 0.000000 0.0 0.0 0.0 Bayview 0.000000 0.0 0.0 0.0 Wista del Mar 0.000000 0.0 0.0 0.0 West Portal 0.023810 0.0 0.0 0.0 Western Addition 0.056604 0.0 0.0 0.0 Westwood Highlands 0.056604 0.0 0.0 0.0	Alamo Square 0.013514 0.0 0.0 0.0 0.0 Anza Vista 0.000000 0.0 0.0 0.0 0.0 Ashbury Heights 0.038462 0.0 0.0 0.0 0.0 Balboa Park 0.000000 0.0 0.0 0.0 0.0 Bayview 0.000000 0.0 0.0 0.0 0.0 Wista del Mar 0.000000 0.0 0.0 0.0 0.0 West Portal 0.023810 0.0 0.0 0.0 0.0 Western Addition 0.000000 0.0 0.0 0.0 0.0 Westwood Highlands 0.056604 0.0 0.0 0.0 0.0	Alamo Square 0.013514 0.0 0.0 0.0 0.0 0.0 0.0 Anza Vista 0.000000 0.0 0.0 0.0 0.0 0.0 0.0 Ashbury Heights 0.038462 0.0 0.0 0.0 0.0 0.0 0.0 Balboa Park 0.000000 0.0 0.0 0.0 0.0 0.0 0.0 Bayview 0.000000 0.0 0.0 0.0 0.0 0.0 0.0 Vista del Mar 0.000000 0.0 0.0 0.0 0.0 0.0 0.0 West Portal 0.023810 0.0 0.0 0.0 0.0 0.0 0.0 Western Addition 0.000000 0.0 0.0 0.0 0.0 0.0 0.0 Westwood Highlands 0.056604 0.0 0.0 0.0 0.0 0.0 0.0 0.0	Neighborhood Studio AIM House Store Acupuncturist Boutique Healer Alamo Square 0.013514 0.0	Neighborhood Studio All House Store Acupuncturist Acupuncturist Boutique Healer Restaurant Alamo Square 0.013514 0.0 0.0 0.0 0.0 0.0 0.0 0.000000 Arabuyry Heights 0.038462 0.0 <td< th=""><th> Alamo Square 0.013514 0.0 0.</th><th> Alamo Square 0.013514 0.0 0.</th><th> Neighborhood Studio Alfo House Store Store Acupuncturist Boutique Healer Restaurant Shelter Wastername Shelter Shelter</th><th> Neighborhood Studio Alfw House Store Acupuncturist Boutique Healer Restaurant Shelter W. Store Restaurant Store Acupuncturist Boutique Healer Restaurant Shelter W. Store Restaurant Alamo Square 0.013514 0.0</th><th> Neighborhood Studio Alfa House Store Store</th></td<>	Alamo Square 0.013514 0.0 0.	Alamo Square 0.013514 0.0 0.	Neighborhood Studio Alfo House Store Store Acupuncturist Boutique Healer Restaurant Shelter Wastername Shelter Shelter	Neighborhood Studio Alfw House Store Acupuncturist Boutique Healer Restaurant Shelter W. Store Restaurant Store Acupuncturist Boutique Healer Restaurant Shelter W. Store Restaurant Alamo Square 0.013514 0.0	Neighborhood Studio Alfa House Store Store

105 rows × 338 columns

sf_grouped

I was curious about the top 10 venues within each neighborhood, so I created a data frame called neighborhoods_venues_sorted to obtain top 10 frequent venues within each neighborhood, following is the head of the neighborhoods_venues_sorted:

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Alamo Square	Bar	Hotel	Café	Record Shop	Sushi Restaurant	Liquor Store	BBQ Joint	Seafood Restaurant	Park	Wine Bar
1	Anza Vista	Café	Cosmetics Shop	Tunnel	Burger Joint	Southern / Soul Food Restaurant	Big Box Store	Mexican Restaurant	Grocery Store	Health & Beauty Service	Liquor Store
2	Ashbury Heights	Café	Bank	Coffee Shop	Dog Run	Supermarket	Middle Eastern Restaurant	Mexican Restaurant	Outdoor Sculpture	Sculpture Garden	Massage Studio
3	Balboa Park	Baseball Field	Café	Breakfast Spot	Bus Station	Flower Shop	BBQ Joint	Asian Restaurant	Light Rail Station	Skate Park	Bus Stop
4	Bayview	Bakery	Southern / Soul Food Restaurant	Café	Mexican Restaurant	Dumpling Restaurant	Pharmacy	Light Rail Station	Home Service	Coffee Shop	Piercing Parlor

neighborhoods_venues_sorted

Part 3: Methodology: K-means Clustering

I used K-means from sklearn.cluster package as my methodology to cluster the neighborhoods. I deleted Neighborhood column in the sf_grouped data frame and then used it to fit my clustering model. I discovered that a 4 clusters model would best cluster my data and provide the most insightful information.

After the aforementioned process, I was able to obtain the following result:

	Neighborhood	lat	Ing	Labels	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10 C
0	Alamo Square	37.776360	-122.434688	1	Bar	Hotel	Café	Record Shop	Sushi Restaurant	Liquor Store	BBQ Joint	Seafood Restaurant	Park	٧
1	Anza Vista	37.780836	-122.443149	1	Café	Cosmetics Shop	Tunnel	Burger Joint	Southern / Soul Food Restaurant	Big Box Store	Mexican Restaurant	Grocery Store	Health & Beauty Service	
2	Ashbury Heights	37.775599	-122.448068	1	Café	Bank	Coffee Shop	Dog Run	Supermarket	Middle Eastern Restaurant	Mexican Restaurant	Outdoor Sculpture	Sculpture Garden	١
3	Balboa Park	37.721427	-122.447547	1	Baseball Field	Café	Breakfast Spot	Bus Station	Flower Shop	BBQ Joint	Asian Restaurant	Light Rail Station	Skate Park	E
4	Bayview	37.728889	-122.392500	1	Bakery	Southern / Soul Food Restaurant	Café	Mexican Restaurant	Dumpling Restaurant	Pharmacy	Light Rail Station	Home Service	Coffee Shop	

101	West Portal	37.741141	-122.465634	1	Italian Restaurant	Wine Bar	Burger Joint	Mexican Restaurant	Chinese Restaurant	Coffee Shop	Pizza Place	Yoga Studio	Diner	В
102	Western Addition	37.779559	-122.429810	1	Liquor Store	Boutique	Grocery Store	Park	Playground	Seafood Restaurant	Farmers Market	Theater	Historic Site	Re
103	Westwood Highlands	37.725726	-122.458199	1	Asian Restaurant	Yoga Studio	Chinese Restaurant	Café	Pharmacy	Coffee Shop	Bubble Tea Shop	Mexican Restaurant	Grocery Store	
104	Westwood Park	37.725726	-122.458199	1	Asian Restaurant	Yoga Studio	Chinese Restaurant	Café	Pharmacy	Coffee Shop	Bubble Tea Shop	Mexican Restaurant	Grocery Store	
105	Butchertown	37.784827	-122.727802	3	Music Venue	Park	Food	Falafel Restaurant	Farmers Market	Fast Food Restaurant	Filipino Restaurant	Fish Market	Flea Market	

106 rows × 14 columns

sf merged

Part 4: Result and Discussion

Cluster 1: Remote Peace

Cluster 1 contains the neighborhoods that are distant from business districts, and these neighborhoods have the most common venues such as parks and trail. Therefore, I would not recommend businesses to start in this cluster, but for people who want to get away from city life, they can visit neighborhoods in this cluster.

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
14	Clarendon Heights	Trail	Park	Art Gallery	Wine Bar	Garden	Reservoir	Monument / Landmark	Playground	Bus Stop	Eye Doctor
31	Forest Hill	Japanese Restaurant	Playground	Park	Hotpot Restaurant	French Restaurant	Football Stadium	Food Truck	Eye Doctor	Falafel Restaurant	Farmers Market
46	Laguna Honda	Trail	Light Rail Station	Jewelry Store	Art Gallery	Hotpot Restaurant	Lake	French Restaurant	Park	Event Space	Bus Stop
58	Merced Manor	Trail	Light Rail Station	Jewelry Store	Art Gallery	Hotpot Restaurant	Lake	French Restaurant	Park	Event Space	Bus Stop
61	Miraloma Park	Bus Stop	Jewelry Store	Park	Trail	Gym	Mountain	Monument / Landmark	Farmers Market	Fast Food Restaurant	Filipino Restaurant
96	Twin Peaks	Trail	Scenic Lookout	Hill	Bus Stop	Bus Station	Reservoir	Food Truck	Filipino Restaurant	Eye Doctor	Fountain

Cluster 2: City Business

This cluster contains the most neighborhoods. Neighborhoods within this clusters are abundant with bars, coffee shops and restaurants. I would suggest a new business to start in these neighborhoods, since they have good business vibes, but new business should conduct research to address imminent competition from other shops. For those who enjoy city life and convenience, this cluster provides the best lifestyles for them.

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Alamo Square	Bar	Hotel	Café	Record Shop	Sushi Restaurant	Liquor Store	BBQ Joint	Seafood Restaurant	Park	Wine Bar
1	Anza Vista	Café	Cosmetics Shop	Tunnel	Burger Joint	Southern / Soul Food Restaurant	Big Box Store	Mexican Restaurant	Grocery Store	Health & Beauty Service	Liquor Store
2	Ashbury Heights	Café	Bank	Coffee Shop	Dog Run	Supermarket	Middle Eastern Restaurant	Mexican Restaurant	Outdoor Sculpture	Sculpture Garden	Massage Studio
3	Balboa Park	Baseball Field	Café	Breakfast Spot	Bus Station	Flower Shop	BBQ Joint	Asian Restaurant	Light Rail Station	Skate Park	Bus Stop
4	Bayview	Bakery	Southern / Soul Food Restaurant	Café	Mexican Restaurant	Dumpling Restaurant	Pharmacy	Light Rail Station	Home Service	Coffee Shop	Piercing Parlor
							***			300	
100	Vista del Mar	Coffee Shop	Café	Hotel	Park	Wine Bar	Theater	Cocktail Bar	Gym	Sushi Restaurant	Juice Bar
101	West Portal	Italian Restaurant	Wine Bar	Burger Joint	Mexican Restaurant	Chinese Restaurant	Coffee Shop	Pizza Place	Yoga Studio	Diner	Bookstore
102	Western Addition	Liquor Store	Boutique	Grocery Store	Park	Playground	Seafood Restaurant	Farmers Market	Theater	Historic Site	German Restaurant
103	Westwood Highlands	Asian Restaurant	Yoga Studio	Chinese Restaurant	Café	Pharmacy	Coffee Shop	Bubble Tea Shop	Mexican Restaurant	Grocery Store	Bank
104	Westwood Park	Asian Restaurant	Yoga Studio	Chinese Restaurant	Café	Pharmacy	Coffee Shop	Bubble Tea Shop	Mexican Restaurant	Grocery Store	Bank

91 rows × 11 columns

Cluster 3: Remote Convenient

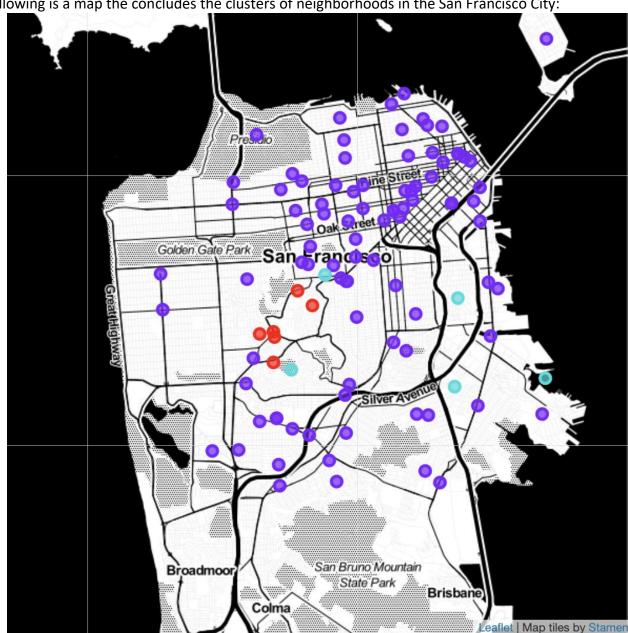
This cluster contains neighborhoods that are distant to business district, but it still provides some level of convenience. For those who want to live close to parks, meanwhile they do not drive to far for some essential stores, this cluster of neighborhoods could be a good option for them. New business can start here if they do not want to face intense level of competitions.

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
38	India Basin	Home Service	Park	Women's Store	Flower Shop	Falafel Restaurant	Farmers Market	Fast Food Restaurant	Filipino Restaurant	Fish Market	Flea Market
55	Lower Nob Hill	Park	Convenience Store	Trail	Road	Dog Run	Scenic Lookout	Shoe Store	Food	Monument / Landmark	Szechuan Restaurant
66	Mount Davidson	Park	Playground	Bus Line	Monument / Landmark	Tree	Falafel Restaurant	Farmers Market	Fast Food Restaurant	Filipino Restaurant	Fish Market
79	Potrero Hill	Grocery Store	Park	Hill	Japanese Restaurant	Gym / Fitness Center	Deli / Bodega	Cosmetics Shop	Convenience Store	Café	Liquor Store
86	Silver Terrace	Grocery Store	Park	Athletics & Sports	Dessert Shop	Soccer Field	Women's Store	Flea Market	Farmers Market	Fast Food Restaurant	Filipino Restaurant

Cluster 4: Fun Mix

This clusters of neighborhoods are made of leisure venues such as music venue, park and markets. For those who want to live close to a mixed range of leisure venues, this cluster is a good choice. New business can also consider this cluster of neighborhoods as a potential choice.

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
10	Cayuga Terrace	Music Venue	Park	Food	Falafel Restaurant	Farmers Market	Fast Food Restaurant	Filipino Restaurant	Fish Market	Flea Market	Flower Shop
59	Midtown Terrace	Music Venue	Park	Food	Falafel Restaurant	Farmers Market	Fast Food Restaurant	Filipino Restaurant	Fish Market	Flea Market	Flower Shop
60	Mid-Market	Music Venue	Park	Food	Falafel Restaurant	Farmers Market	Fast Food Restaurant	Filipino Restaurant	Fish Market	Flea Market	Flower Shop
105	Butchertown	Music Venue	Park	Food	Falafel	Farmers	Fast Food	Filipino	Fish Market	Flea Market	Flower Shop



Following is a map the concludes the clusters of neighborhoods in the San Francisco City:

Part 5: Conclusion

Based on the API and neighborhood data, I used the K-means clusters to divide the neighborhoods in San Francisco into four clusters, which are Remote Peace, City Business, Remote Convenient and Mix Fun. Businesses and city resident can use this report as a resource to choose the neighborhoods according to their need.