

# Final Report: The Battle of the Neighborhoods

Christopher González Delgado

# 1 Introduction. Objectives and target

Physical exercise and the desire to have a fit, muscular and attractive body are a trend in our society. Therefore, it would not be unreasonable for a person to see an ideal business opportunity in setting up a gym. And although the idea is good, there is an unknown question that we must solve: where should we place the gym so that it is a successful site? Through this project you will be able to observe the methodology to locate in a good area the business of the 21st century: the fitness centers.

The objective of this project is to find the ideal areas to open a gym in Toronto taking into account some characteristics of the area. The audience that could show the most interest consists of investors looking for a safe bet or entrepreneurs who want to start their business.

## 2 Data description and extraction

Firstly, we have a database (obtained from wikipedia) that offers us the postal code, the boroughs and the neighborhoods, and secondly, another database that consists of the coordinates of these places. On the other hand, the Python package "geopy" has been used to obtain the coordinates of Toronto on the map, while the Foursquare platform has been used to obtain information about the venues in Toronto.

Postal Code ↕	Borough ↕	Neighbourhood
M1A	Not assigned	Not assigned
M2A	Not assigned	Not assigned
M3A	North York	Parkwoods
M4A	North York	Victoria Village
M5A	Downtown Toronto	Regent Park, Harbourfront
M6A	North York	Lawrence Manor, Lawrence Heights
M7A	Downtown Toronto	Queen's Park, Ontario Provincial Government
M8A	Not assigned	Not assigned
M9A	Etobicoke	Islington Avenue, Humber Valley Village

Figure 1: Example of the Wikipedia dataset

## 3 Methodology

### 3.1 Understanding the problem

Obviously the main problem is: where should we place the gym? But to complete this question we must also add certain conditions: **where should we place the gym if we take into account the other shops and establishments in each neighborhood?**

### 3.2 Preparation of the data and processing

The data we have obtained, as seen in Figure 1, is dirty i.e. it has some unassigned variables whose value we cannot intuit or calculate. Therefore, first of all we are going to delete all the data whose "Borough" has the value "Not assigned". Once this is done, we can see that there are several postal codes that appear more than once and with different neighborhoods, so we

will proceed to merge these neighborhoods in the same row for each postal code. Finally, we will focus only on those boroughs that explicitly have the word "Toronto" in their name.

```
In [2]: df = pd.read_html('https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M')
df = df[0]
dff = df.drop( df[(df['Borough']!= 'Not assigned')].index)
dff = dff.reset_index(drop=True) #Let's reset the index

In [8]: dff.columns=(['Postal Code', 'Borough', 'Neighborhood'])
dff.head()
dff3= dff.groupby(['Postal Code', 'Borough'])['Neighborhood'].apply(list).apply(lambda x: ", ".join(x)).to_frame()
dff3 = dff3.reset_index() #Let's add the index to each row
```

Figure 2: Pre-processing the data

As a final step in the creation of the "useful" dataset, we merge both dataset: the one that contains the neighborhoods and the one that contains the geographical coordinates.

### 3.3 Working with the data. Results

The procedure that has been carried out has been as follows: First of all we have obtained the list of venues in toronto.

```
In [13]: def getNearbyVenues(names, latitudes, longitudes, radius=500):
venues_list=[]
for name, lat, lng in zip(names, latitudes, longitudes):
    print(name)
    # create the API request URL
    url = 'https://api.foursquare.com/v2/venues/explore?client_id={}&client_secret={}&v={}&ll={}&radius={}&limit={}&nearby={}'
    CLIENT_ID,
    CLIENT_SECRET,
    VERSION,
    lat,
    lng,
    radius,
    LIMIT)

    # make the GET request
    results = requests.get(url).json()[ "response" ][ 'groups' ][ 0 ][ 'items' ]
    # return only relevant information for each nearby venue
    venues_list.append([
        name,
        lat,
        lng,
        v[ 'venue' ][ 'name' ],
        v[ 'venue' ][ 'location' ][ 'lat' ],
        v[ 'venue' ][ 'location' ][ 'lng' ],
        v[ 'venue' ][ 'categories' ][ 0 ][ 'name' ] for v in results])
nearby_venues = pd.DataFrame([item for venue_list in venues_list for item in venue_list])
nearby_venues.columns = [ 'Neighborhood',
    'Neighborhood Latitude',
    'Neighborhood Longitude',
    'Venue',
    'Venue Latitude',
    'Venue Longitude',
    'Venue Category' ]

return(nearby_venues)

In [14]: LIMIT=50
toronto_venues = getNearbyVenues(names=DFF['Neighborhood'],
    latitudes=DFF['Latitude'],
    longitudes=DFF['Longitude']
)
```

Figure 3: Obtaining Toronto venues

From these data, it has been possible to search for the neighborhoods that do not have any (commonly) gym or fitness center in their surroundings (we are interested in the fact that they do not have gyms, so ours will draw the attention of people who live near that area ). On the other hand, it has been taken into account that it is an area with many cafes and hotels. The first is evident: coffee is a good sports supplement and it is common to drink coffee only before training, in addition to being common to take some food or protein shake after finishing workouts. And the second reason is because when people go on vacation they are likely to

want to maintain moderate physical activity: if we can locate a gym near a hotel we can have a fairly constant flow of customers. Once these conditions were established, an analysis of the data has been carried out and it has been concluded that some of the neighborhoods that meet these conditions are the ones shown below:

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 Garden District, Ryerson	Coffee Shop	Café	Clothing Store	Cosmetics Shop	Ramen Restaurant	Tea Room	Theater	Bookstore	Fast Food Restaurant	Japanese Restaurant
1 Toronto Dominion Centre, Design Exchange	Coffee Shop	Café	Seafood Restaurant	Hotel	Japanese Restaurant	Restaurant	Beer Bar	Bakery	Gym / Fitness Center	Deli / Bodega
2 Kensington Market, Chinatown, Grange Park	Café	Mexican Restaurant	Vegetarian / Vegan Restaurant	Coffee Shop	Burger Joint	Vietnamese Restaurant	Bar	Pizza Place	Dessert Shop	Record Shop
3 Runnymede, Swansea	Café	Coffee Shop	Pizza Place	Pub	Sushi Restaurant	Italian Restaurant	Yoga Studio	Bookstore	Bar	Smoothie Shop

Figure 4: Suitable neighborhoods to place a Fitness Center

Finally, a map has been constructed using Folium to visualize the areas where it would be ideal to set up a gym:

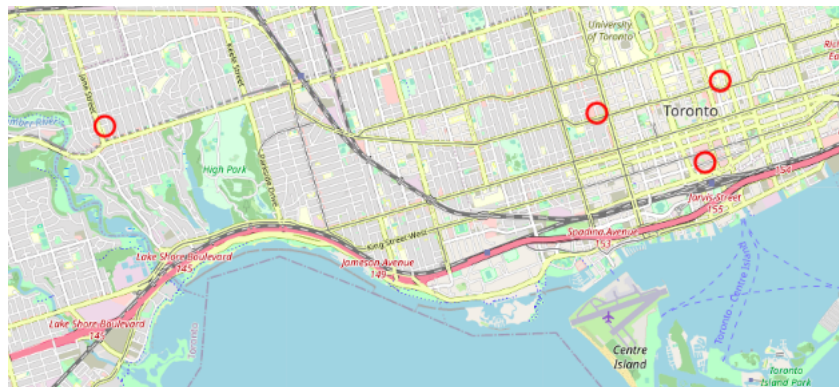


Figure 5: Map showing the four places to open a Fitness Center

## 4 Discussion and conclusions

Through this project, it has been possible to demonstrate that using powerful data science tools and using a simple algorithm, really powerful results can be obtained. Specifically, we have seen that there are (at least) 4 areas in which it would be ideal to open a gym according to the conditions, which are: cafes and hotels nearby.

Much deeper analysis could be done using other variables that play a vital role in all of this. For example, including the average income of the neighborhoods or the average age. If we include the rent we can focus on a more affluent public that is willing to pay higher amounts for the service; Taking into account the age of potential clients, we can look for a young audience as they are interested in obtaining a muscular and aesthetic body.

## 5 References

- List of Postal Codes of Canada: [https://en.wikipedia.org/wiki/List\\_of\\_postal\\_codes\\_of\\_Canada:\\_M](https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M)
- Geographical Coordinates of the Neighborhoods: [https://cocl.us/Geospatial\\_data](https://cocl.us/Geospatial_data)