

# **Battle Of The Neighborhoods**

## **Worship Site Planning**

## **Table of Contents**

Introduction  
Data  
Methodology  
Results  
Discussion  
Conclusion  
...

# **The Battle of the Neighborhoods**

## **Spiritual Site Planning in the 21<sup>st</sup> Century**

### **Introduction**

This is a proposal that is targeted to Church / Houses of Worship planners. It is intended to provide first level information about the current places of worship: information that can be used to identify a candidate location for installing a new community-based religious outreach. To do this, it is necessary to come to a balance between honoring existing ministries, while also being aware of the need for a different perspective to serve a larger and more diverse audience. Specifically, there are two principles that originated in the Christian Bible which are informative in explaining the challenge of being true in selecting a site location.

Consider the following that comes from the Bible, in Romans 15:20-21:

*Yea, so have I strived to preach the gospel, not where Christ was named, lest I should build upon another man's foundation: But as it is written, To whom he was not spoken of, they shall see: and they that have not heard shall understand.*

This challenges the planner to select a site that has a need for the theological emphasis of the house of worship being designed. It is not necessary that there be no existing structures of that theological persuasion; but it does indicate that there should be some level of additional need, where there is already some support. Of course, ideally, the planner will locate a 'virgin' site.

Also consider the following that comes from the Bible, in the first epistle of Paul the apostle to the Corinthians 9:19-23:

*For though I be free from all men, yet have I made myself servant unto all, that I might gain the more. And unto the Jews I became as a Jew, that I might gain the Jews; to them that are under the law, as under the law, that I might gain them that are under the law; To them that are without law, as without law, (being not without law to God, but under the law to Christ,) that I might gain them that are without law. To the weak became I as weak, that I might gain the weak: I am made all things to all men, that I might by all means save some. And this I do for the gospel's sake, that I might be partaker thereof with you.*

This challenges the planner to select a location that has more than the single theological focus of the site plan. This is necessary because one of the strongest purposes of building a house of worship is to interact with others who have a mind to serve God. This means that there is a necessity for communication of the message that brings the overseers of the house of worship (pastor, priests, bishops, etc.) to this location. This is the process that includes proselytizing. So, in the surrounding environment of the proposed house of worship, there needs to be other type of theological persuasions (temples, prayer rooms, etc.)

This is the challenge that leads us to use modern data analysis and machine learning methods, to extract insights from existing information about boroughs and neighborhoods. For purposes of illustration, we will look at the Toronto, Canada area.

-----

## Spiritual Site Planning in the 21<sup>st</sup> Century

### Data

To recap, from the introduction, we stated the following: this is the challenge that leads us to use modern data analysis and machine learning methods, to extract insights from existing information about boroughs and neighborhoods.

In this study we will manipulate the following resources for location information, as indicated:

1. Scrape the following Wikipedia page, [https://en.wikipedia.org/wiki/List\\_of\\_postal\\_codes\\_of\\_Canada:\\_M](https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M), in order to obtain the data that is in the table of postal codes and to transform the data into a pandas dataframe. The dataframe will consist of three columns: Postal Code, Borough, and Neighbourhood. For example:

	<b>Postal Code</b>	<b>Borough</b>	<b>Neighbourhood</b>
4	M5A	Downtown Toronto	Regent Park, Harbourfront

2. Additional information will be obtained by utilizing the following link to a csv file that has the geographical coordinates of each postal code: [http://cocl.us/Geospatial\\_data](http://cocl.us/Geospatial_data). An example of the information that is needed from there is shown below:

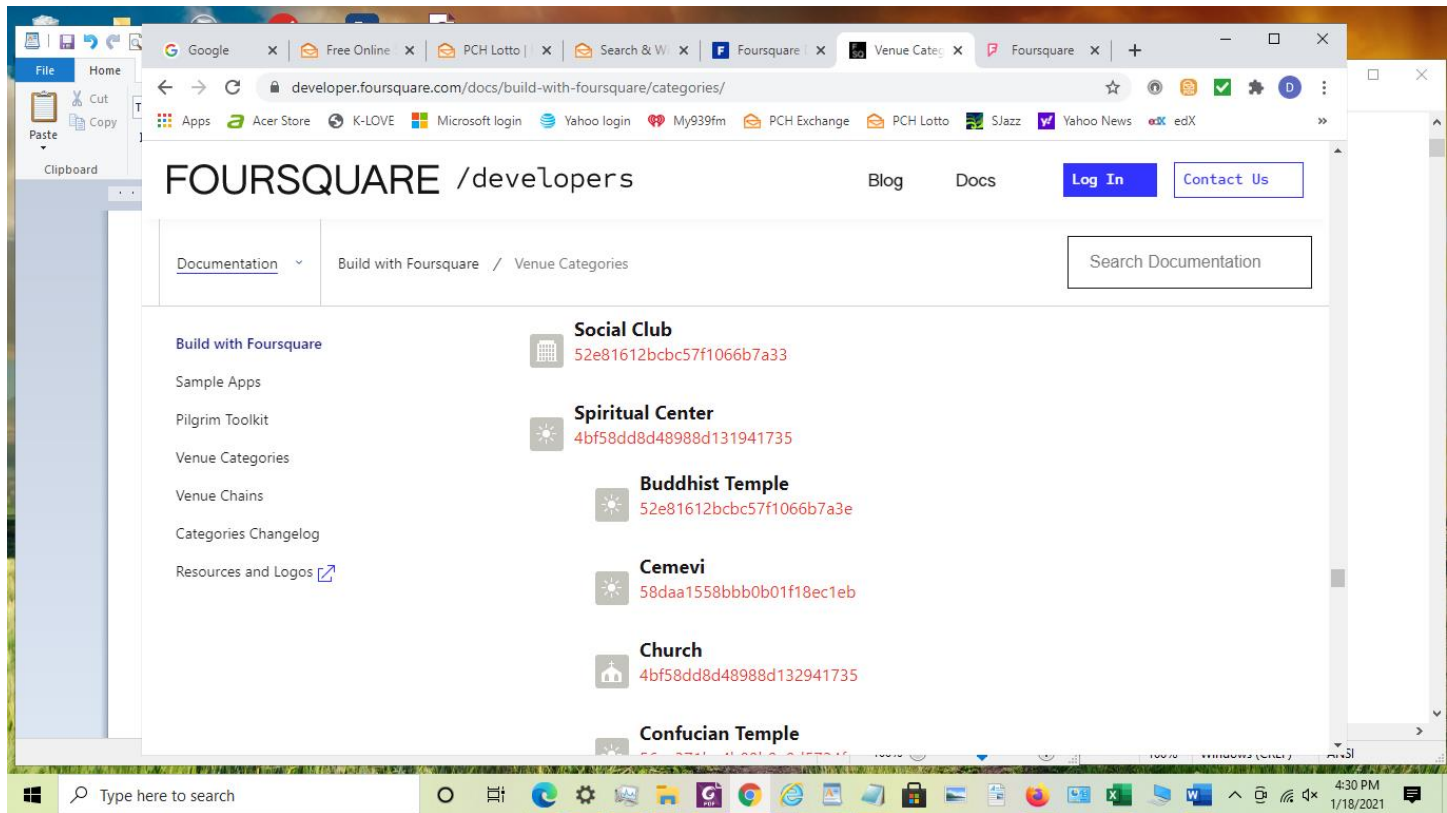
	<b>Postal Code</b>	<b>Latitude</b>	<b>Longitude</b>
2	M1E	43.763573	-79.188711

We will also need information on the existing houses of worship. This type of information is contained in what are known as venues. One of the premier locations for obtaining borough and neighborhood venue information is Foursquare. This is the place where we will search for the necessary information about existing houses of worship. The level of data that is needed include more than just the borough and neighborhood of the house of worship. We must also be able to extract the theological focus of the houses that are contained in those location.

Once we have obtained this collection, we will be able to apply machine learning to order the information in ways that allow us to see patterns of need in the various locations. That need will be based on the specific type of house of worship for which we are planning.

On the following page, there is a screenshot of some of the information that is provided in the Foursquare offering.

## Foursquare Sample Categories



We will have more to say about the use of this information in the Methodology section of the report.

-----

## Spiritual Site Planning in the 21<sup>st</sup> Century

### Methodology

To perform this analysis, the first thing that must be done is to prepare a pandas dataframe of Toronto entities and their addresses, with their associated latitude and longitude values. To get this information we gather the following columns of data:

1. From [https://en.wikipedia.org/wiki/List\\_of\\_postal\\_codes\\_of\\_Canada:\\_M](https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M), we obtain the following data that is in the list of postal codes:

	Postal Code	Borough	Neighbourhood
4	M5A	Downtown Toronto	Regent Park, Harbourfront

2. From [http://cocl.us/Geospatial\\_data](http://cocl.us/Geospatial_data) we obtain the geographical coordinates of each postal code, for example:

	Postal Code	Latitude	Longitude
2	M1E	43.763573	-79.188711

To use the geographical coordinates of each postal code, we need to remove the “Not Assigned” Boroughs, since the “Not Assigned” postal codes are not in the listing.

Now, we join two collections of information to make the following type of rows of a dataframe:

	Postal Code	Borough	Neighbourhood	Latitude	Longitude
4	M5A	Downtown Toronto	Regent Park, Harbourfront	43.763573	-79.188711

Once the above join is done, we drop the postal codes from the pandas dataframe, since we will be working primarily with Boroughs and Neighbourhoods. The following is an example of the resultant dataframe, for all Canada Boroughs.

	Borough	Neighbourhood	Latitude	Longitude
4	Downtown Toronto	Regent Park, Harbourfront	43.763573	-79.188711

From the above dataframe, we extract only those rows that contain the word “Toronto” in the Borough. This is our candidate set for further processing. This is the set that will be parsed through an Internet offering known as Foursquare. A brief description of that offering is included below, from the Foursquare.com “About Us”.

*“It all starts with the data - and ours is unmatched in quality and depth. We combine the rich attributes of over 105 million global points-of-interest with the understanding of human movement from over 500 million devices, including a persistent first-party panel.”*

Now, we use the Foursquare API to explore neighborhoods in Toronto. We use the explore function to get certain spiritual venue categories in each neighborhood.

Then, we take the venues list, and group the neighborhoods into clusters: to do this, we use the k-means clustering algorithm to complete this task. As described on the Internet: *“This algorithm is an iterative algorithm that partitions the dataset according to their features into K number of predefined non-overlapping distinct clusters or subgroups. It makes the data points of inter clusters as similar as possible and also tries to keep the clusters as far as possible.”*

This tool allows us to order locations by their predominant houses of worship, then produce clusters of neighborhoods that may benefit from a particular offering of religious outreach. By evaluating these clusters, we will be able to identify places that meet the two principles of religious site planting:

1. *“Yea, so have I strived to preach the gospel, not where Christ was named, lest I should build upon another man's foundation.”*
2. *“For though I be free from all men, yet have I made myself servant unto all, that I might gain the more.” . . . “I am made all things to all men, that I might by all means save some.”*

We only select the top ten unique sites for further evaluation. Please note that some of the neighborhoods will not have any spiritual sites that match our top ten selection criteria; so, first we remove neighborhoods that do not have any representative spiritual sites. This is done by removing any rows that have NaN in the label (‘Cluster labels’) that identifies the collection to which each neighborhood belongs.

Finally, we will use the Folium library to visualize the neighborhoods and their emerging clusters.

We deployed all these tools in a python notebook that we push to GitHub. We will have more to say about the deliverables from this notebook in the following section: Results.

-----

# Spiritual Site Planning in the 21<sup>st</sup> Century

## Results

The following shows the clusters of neighborhoods, in two images.

-----

Firstly, here are screen prints of all or a portion of the cluster collections that are notable as potential sites.

### Cluster One

The screenshot displays a Jupyter Notebook interface within the IBM Cloud Pak for Data environment. The notebook is titled "Church Planting in Toronto - Cour...". The code cell shows the selection of Cluster 1 from a dataset named "Toronto\_merged". The output is a table with 11 columns: "Neighbourhood", "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", and "10th Most Common Venue". Two rows of data are visible, corresponding to neighborhoods 31 and 37.

	Neighbourhood	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
31	Summerhill West, Rathnelly, South Hill, Forest...	Spiritual Center	Church	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
37	Church and Wellesley	Gay Bar	Church	Residential Building (Apartment / Condo)	Breakfast Spot	Bookstore	Spiritual Center	Park	Office	Music Venue	Monument / Landmark

### Cluster Two



IBM Cloud Pak for Data

Projects / Duane Andry Coursera / Church Planting in Toronto - Cour...

AutoAI Experiments has a new beta feature that supports the joining of multiple data sets into a single data source, with richer feature creation and transformation capabilities. [Learn more.](#)

File Edit View Insert Cell Kernel Help Not Trusted | Python 3.7

```
In [54]: # Cluster 2
Toronto_merged.loc[Toronto_merged['Cluster Labels'] == 1, Toronto_merged.columns[[1] + list(range(5, Toronto_merged.shape[1]))]]
```

Out[54]:

	Neighbourhood	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
1	Queen's Park, Ontario Provincial Government	Church	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
2	Garden District, Ryerson	Church	Office	Spiritual Center	Residential Building (Apartment / Condo)	Park	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
3	St. James Town	Church	Residential Building (Apartment / Condo)	Spiritual Center	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
4	The Beaches	Church	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
6	Central Bay Street	Church	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space

Church Planting...ipynb

## Cluster Three

IBM Cloud Pak for Data

Projects / Duane Andry Coursera / Church Planting in Toronto - Cour...

AutoAI Experiments has a new beta feature that supports the joining of multiple data sets into a single data source, with richer feature creation and transformation capabilities. [Learn more.](#)

File Edit View Insert Cell Kernel Help Not Trusted | Python 3.7

```
In [55]: # Cluster 3
Toronto_merged.loc[Toronto_merged['Cluster Labels'] == 2, Toronto_merged.columns[[1] + list(range(5, Toronto_merged.shape[1]))]]
```

Out[55]:

	Neighbourhood	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
11	Little Portugal, Trinity	Event Space	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Church

```
In [56]: # Cluster 4
Toronto_merged.loc[Toronto_merged['Cluster Labels'] == 3, Toronto_merged.columns[[1] + list(range(5, Toronto_merged.shape[1]))]]
```

Out[56]:

	Neighbourhood	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	Harbourfront East, Union Station, Toronto Islands	Monument / Landmark	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Intersection	Gay Bar	Event Space	Church

Church Planting...ipynb

## Cluster Four

The screenshot shows the IBM Cloud Pak for Data interface. The top navigation bar includes 'IBM Cloud Pak for Data', 'All', a search bar, and 'Upgrade'. The main header displays 'Projects / Duane Andry Coursera / Church Planting in Toronto - Cour...'. A blue announcement banner states: 'AutoAI Experiments has a new beta feature that supports the joining of multiple data sets into a single data source, with richer feature creation and transformation capabilities. [Learn more.](#)'

The interface shows a Jupyter Notebook with the following code and output:

```
In [56]: # Cluster 4
Toronto_merged.loc[Toronto_merged['Cluster Labels'] == 3, Toronto_merged.columns[[1] + list(range(5, Toronto_merged.shape[1]))]]
```

Out[56]:

	Neighbourhood	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	Harbourfront East, Union Station, Toronto Islands	Monument / Landmark	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Intersection	Gay Bar	Event Space	Church

The notebook continues with:

```
In [57]: # Cluster 5
Toronto_merged.loc[Toronto_merged['Cluster Labels'] == 4, Toronto_merged.columns[[1] + list(range(5, Toronto_merged.shape[1]))]]
```

Out[57]:

	Neighbourhood	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	Regent Park, Harbourfront	Church	Park	Event Space	Spiritual Center	Residential Building (Apartment / Condo)	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar

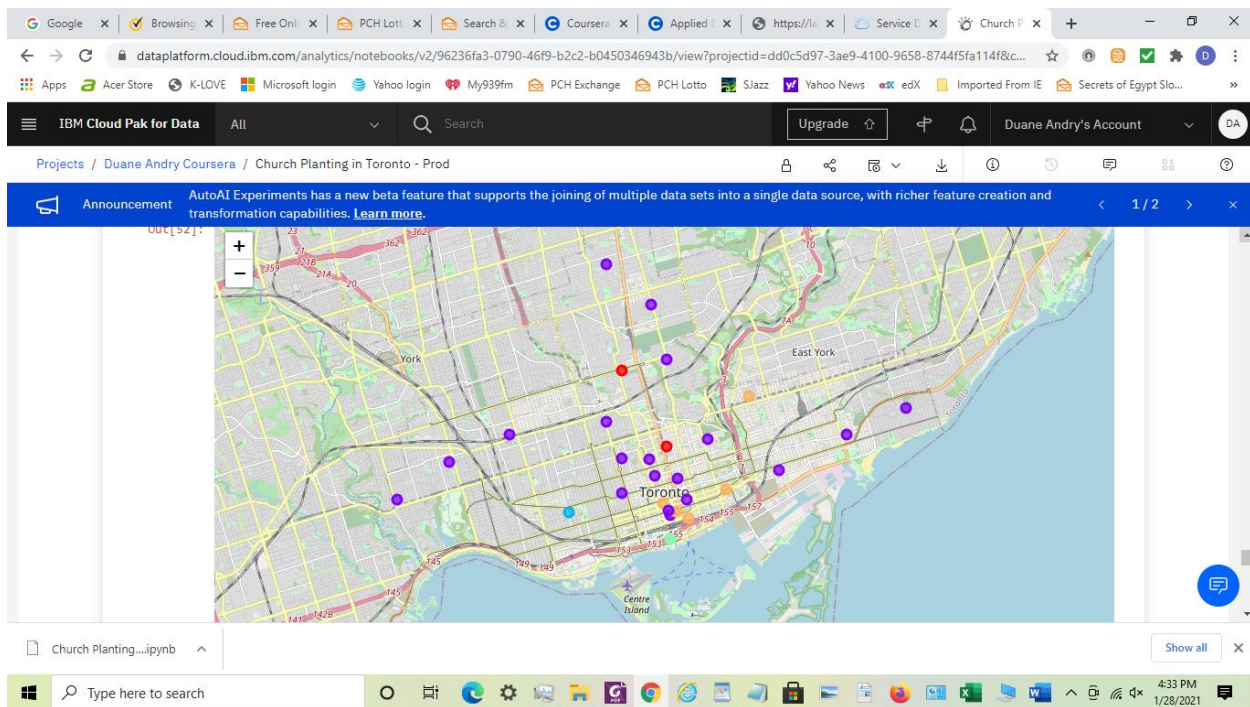
## Cluster Five

The screenshot shows the IBM Cloud Pak for Data interface, similar to the previous one, but displaying the output for Cluster 5. The code and announcement banner are identical.

Out[57]:

	Neighbourhood	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	Regent Park, Harbourfront	Church	Park	Event Space	Spiritual Center	Residential Building (Apartment / Condo)	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar
8	Richmond, Adelaide, King	Church	Office	Spiritual Center	Residential Building (Apartment / Condo)	Park	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
12	The Danforth West, Riverdale	Church	Music Venue	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Monument / Landmark	Intersection	Gay Bar	Event Space
16	Commerce Court, Victoria Hotel	Church	Intersection	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Gay Bar	Event Space
34	Stn APO Boxes	Intersection	Church	Spiritual Center	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Gay Bar	Event Space

Shown below, there is a visual representation of the cluster collections:



Also of interest is this collection of neighborhoods that do not contain any representative venues, and that, therefore, do not fit in any of the collections. We are including them here for the sake of completion of the report on boroughs and neighborhoods in Toronto.

In [43]:

```
Toronto_merged1 = Toronto_merged[Toronto_merged['Cluster Labels'].isna()]
Toronto_merged1
```

Out [43]:

	Borough	Neighbourhood	Latitude	Longitude	Cluster 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
5	Downtown Toronto	Berczy Park	43.644771	-79.373306	NaN	NaN	NaN	NaN	NaN	NaN	NaN
7	Downtown Toronto	Christie	43.669542	-79.422564	NaN	NaN	NaN	NaN	NaN	NaN	NaN
14	West Toronto	Brockton, Parkdale Village, Exhibition Place	43.636847	-79.428191	NaN	NaN	NaN	NaN	NaN	NaN	NaN
18	Central Toronto	Lawrence Park	43.728020	-79.388790	NaN	NaN	NaN	NaN	NaN	NaN	NaN
19	Central Toronto	Roselawn	43.711695	-79.416936	NaN	NaN	NaN	NaN	NaN	NaN	NaN
20	Central Toronto	Davisville North	43.712751	-79.390197	NaN	NaN	NaN	NaN	NaN	NaN	NaN
21	Central Toronto	Forest Hill North & West, Forest Hill Road Park	43.696948	-79.411307	NaN	NaN	NaN	NaN	NaN	NaN	NaN
	West	Parkdale									

14	West Toronto	Parkdale Village, Exhibition Place	43.636847	-79.428191	NaN	NaN	NaN	NaN	NaN	NaN	NaN
18	Central Toronto	Lawrence Park	43.728020	-79.388790	NaN	NaN	NaN	NaN	NaN	NaN	NaN
19	Central Toronto	Roselawn	43.711695	-79.416936	NaN	NaN	NaN	NaN	NaN	NaN	NaN
20	Central Toronto	Davisville North	43.712751	-79.390197	NaN	NaN	NaN	NaN	NaN	NaN	NaN
21	Central Toronto	Forest Hill North & West, Forest Hill Road Park	43.696948	-79.411307	NaN	NaN	NaN	NaN	NaN	NaN	NaN
25	West Toronto	Parkdale, Roncesvalles	43.648960	-79.456325	NaN	NaN	NaN	NaN	NaN	NaN	NaN
32	Downtown Toronto	CN Tower, King and Spadina, Railway Lands, Har...	43.628947	-79.394420	NaN	NaN	NaN	NaN	NaN	NaN	NaN
33	Downtown Toronto	Rosedale	43.679563	-79.377529	NaN	NaN	NaN	NaN	NaN	NaN	NaN
38	East Toronto	Business reply mail Processing Centre, South C...	43.662744	-79.321558	NaN	NaN	NaN	NaN	NaN	NaN	NaN

-----

Note: for the partial collections, their full depth can be seen in the notebook associated with this report.

-----

With this information, we can take a good look at the results, in the neighborhoods clustering of the python notebook, to perform the two-fold evaluation of the sites that will give us a candidate location for our religious offering. This, we will do in the next section: Analysis.

-----

# Spiritual Site Planning in the 21<sup>st</sup> Century

## Discussion

To satisfy principle one, we will evaluate the clustering of neighborhood to determine a cluster that does not have either a predominance or a prominence of our intended site of worship. For purposes of our evaluation, here, we will look for a site that would support the planting of a Christian church.

If more than one location satisfies principle one; then, to satisfy principle two, we will evaluate the selected cluster individually, from the collection of clusters, to locate a neighborhood that has the greatest breadth of other offering of religious outreach.

Considering the Results itemized in the prior section, as applied in its use for placement of a Christian church, we deselect cluster four (4), since its label seems to indicate that it is not a residential section of Toronto.

From the clusters that remain, we deselect ones that have a predominance or prominence of “Church”; for instance, where it is in the first position. These are clusters two (2) and five (5).

From the remaining clusters – one (1) and three (3) -- we have chosen from cluster one (1) and the neighborhood on row 37:

The screenshot shows a Jupyter notebook interface within the IBM Cloud Pak for Data environment. The notebook is titled "Church Planting in Toronto - Cour...". It displays two code cells and their corresponding output tables.

**Cluster 1 Output:**

	Neighbourhood	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
31	Summerhill West, Rathnelly, South Hill, Forest...	Spiritual Center	Church	Residential Building (Apartment / Condo)	Park	Office	Music Venue	Monument / Landmark	Intersection	Gay Bar	Event Space
37	Church and Wellesley	Gay Bar	Church	Residential Building (Apartment / Condo)	Breakfast Spot	Bookstore	Spiritual Center	Park	Office	Music Venue	Monument / Landmark

**Cluster 2 Output:**

	Neighbourhood	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
--	---------------	-----------------------	-----------------------	-----------------------	-----------------------	-----------------------	-----------------------	-----------------------	-----------------------	-----------------------	------------------------

In this neighborhood, we see that “Church” is not the first venue identified, but, nonetheless, there are other locations that are churches. Also in this neighborhood are venues that provide for further outreach because they are not churches. This meets the two-fold dynamic of religious site planning, as it pertains to a Christian church, which we described in the reasoning for site selection. Also, as we look further into the cluster, we see that there is another neighborhood that allow for potential proselytizing in the Christian message. This is the concept of sharing that is an integral part of being a Christian outreach.

## **Spiritual Site Planning in the 21<sup>st</sup> Century**

### **Conclusion**

As we have shown, machine learning allows for a robust evaluation of neighborhood components that are germane to church planting. In this report, we have covered the first stage of church planting: site selection. The use of machine learning produces a reusable model that will serve for revisiting the factors that lead us to a particular location, as opposed to other locations.

Machine learning, as done in this study, shows us the cluster of neighborhoods that can benefit from the offering of services in our current portfolio of outreach potential. Thus, it serves to allow the maximization of assets, to produce the best outcomes for the impacts that we can make. Moreover, there will be opportunities to evolve our portfolio, to include other services. In that respect, it is an organismic approach to worship offering.

We also need to consider the condition where there is no need for evolution in our portfolio of services. This does not mean that we can discard the tool when we have maximized the provision of service. Another factor that must be considered is neighborhood change itself. This tool can be utilized, from time to time to support the redeployment of resources as needed, as the dynamics of the neighborhood cluster changes. For instance, at such times as additional organizations enter the cluster of neighborhoods, or entities remove themselves from service in this area, we can reevaluate the environment without changing this tool that originally brought us to this place of service.

Reuse of the tool is the power of employing machine learning to the environment, to provide insight into our place (or not) of service therein.