Segmenting and Clustering Toronto neighbourhoods(2) (3).ipynb (/github/Kenshinhuang/Applied-Data-Science-Capstone/tree/master/Segmenting Toronto neighbourhoods(2) (3

Segmenting and Clustering Toronto Neighbourhoods

Objective: Segment and cluster the Toronto neighbourhoods based on post codes

Download the table of post codes for neighbourhoods in Toronto from Wikipedia. https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada: M (https://en.wiki/List_of_Dostal_codes_of_Canada:

I have Excel 2016, which can hold 1,048,576 rows. The post code table has only 287 rows. I tried using BeautifulSoup first. But when I was half-way through and dealing with all those issues of cleaning the data, I realized that for such a small table, directlying downloading it into Excel, then uploading it into Jupyter Notebook will be much easier and faster as well. So, I scraped all my BeautifulSoup codes.

```
In [1]:
               import pandas as pd
               postcode_df=pd.read_excel('Toronto Post Codes.xlsx')
               postcode_df.head()
Out[1]:
                  Postcode
                                  Borough Neighbourhood
                      M1A
                               Not assigned
                                              Not assigned
                      M2A
                                              Not assigned
                               Not assigned
                      МЗА
                                 North York
                                               Parkwoods
                      M4A
                                 North York
                                             Victoria Village
                      M5A Downtown Toronto
                                              Harbourfront
In [2]:
               # size of the table
               postcode_df.shape
Out[2]:
               (287, 3)
In [3]:
               # number of unique Boroughs
               import numpy as np
               postcode df["Borough"].value counts()
Out[3]:
              Not assigned
              Etobicoke
                                    45
              North York
                                    38
              Downtown Toronto
                                    37
              Scarborough
                                    37
              Central Toronto
                                    17
              West Toronto
                                    13
              York
                                     9
              East Toronto
                                     7
              East York
              Mississauga
              Name: Borough, dtype: int64
```

Okay. So there are 77 "Not assigned" in the Borough column. I could use drop.duplicates() to drop those "Not assigned". However, that will drop the other boroughs as well, since there are boroughs that have more than one neighbourhoods assigned to it. So, drop by "Borough" won't work.

Let's check the Neighbourhood column. The neighbourhoods should be unique, except for the "Not assigned".

```
In [4]: postcode_df["Neighbourhood"].value_counts()
```

 Out[4]:
 Not assigned
 77

 St. James Town
 2

 Runnymede
 2

 Yorkville
 1

 Glencairn
 1

 Exhibition Place
 1

 Morningside
 1

 Parkdale
 1

 Bathurst Manor
 1

Scarborough Village West

Name: Neighbourhood, Length: 209, dtype: int64

All the neighbourhoods are unique, except for 1) the "Not assigned" and 2) "Runnymede" and "St. James Town" where there are two copies.

In [5]: #Remove the "Not assigned" from the Borough column.
postcode_df.drop_duplicates(subset="Neighbourhood",keep=False, inplace=True)
postcode_df

Out[5]:

Neighbourhood	Borough	Postcode	
Parkwoods	North York	МЗА	2
Victoria Village	North York	M4A	3
Harbourfront	Downtown Toronto	M5A	4
Lawrence Heights	North York	M6A	5
Lawrence Manor	North York	M6A	6
Kingsway Park South West	Etobicoke	M8Z	281
Mimico NW	Etobicoke	M8Z	282
The Queensway West	Etobicoke	M8Z	283
Royal York South West	Etobicoke	M8Z	284
South of Bloor	Etobicoke	M8Z	285

206 rows × 3 columns

All the "Not assigned" in the Neighbourhood column has been dropped. However, since the parameter "keep" was set to False. The process above dropped the two neighbourhoods "Runnemede" and "St. James Town" as well. But I would like to keep those two neighbourhoods, since each copy belong to two different boroughs.

Put the two neighbourhoods back.

In [6]: append1=pd.DataFrame({"Postcode":["M5C","M6N","M6S","M4X"],"Borough":["Downtown Toronto","York","West Toronto","Downtown Toronto", one postcode_df.append(append1,ignore_index=False)

Out[6]:

Neighbourhood	Borough	Postcode	
Parkwoods	North York	МЗА	2
Victoria Village	North York	M4A	3
Harbourfront	Downtown Toronto	M5A	4
Lawrence Heights	North York	M6A	5
Lawrence Manor	North York	M6A	6
South of Bloor	Etobicoke	M8Z	285
St. James Town	Downtown Toronto	M5C	0
Runnymede	York	M6N	1
Runnymede	West Toronto	M6S	2
St. James Town	Downtown Toronto	M4X	3

210 rows × 3 columns

In [12]: postcode_df=postcode_df.append(append1,ignore_index=False)

```
In [13]: print("The cleaned table has", len(postcode_df["Postcode"]), "row")
```

The cleaned table has 210 row

As shown above, the original table has 287 rows in total and 77 "Not assigned" in the Borough column. After the 77 "Not assigned" rows are dropped, the cleaned table should have 287-77=210 rows. The result of the above code block shows 210 rows.

Let's confirm that there is no more "Not assigned" in Borough column.

```
In [14]: postcode_df["Borough"].value_counts()
```

Out[14]:

Etobicoke North York Downtown Toronto 37 Scarborough 37 Central Toronto 17 West Toronto 13 York East Toronto East York 6 Mississauga Name: Borough, dtype: int64

Cool. There is no more "Not assigned" in the Borough column.

The table is cleaned now. I can start working with it. The next thing I am going to do is to put the neighbourhoods that have the same postcode in the same row. Let's see how many unique postcodes are in the cleaned table.

```
In [16]: postcode_df["Postcode"].value_counts()
```

Out[16]:

```
M8Y
       8
M9V
M5V
M8Z
       5
м9в
      5
M4A
      1
M4N
M5G
       1
M5N
       1
M6G
```

Name: Postcode, Length: 103, dtype: int64

There are 103 unique postcodes. Most of the postcodes has only 1 neighbourhoods associated with it. However, postcodeS "M8Y" and "M9V" have 8 neighbourhoods associated with it, "M5V" has 7, "M4V" and "M8Z" have 5. I need to concatenate all those neighbourhoods under the postcode they are associated with, in one row.

```
In [17]: postcode_df=pd.DataFrame(postcode_df.groupby(["Postcode","Borough"])["Neighbourhood"].apply(lambda x: ','.join(x)))
postcode_df
```

Out[17]:

Neighbourhood

	Borough	Postcode
Rouge,Malvern	Scarborough	M1B
Highland Creek,Rouge Hill,Port Union	Scarborough	M1C
Guildwood, Morningside, West Hill	Scarborough	M1E
Woburn	Scarborough	M1G
Cedarbrae	Scarborough	M1H
Weston	York	M9N
Westmount	Etobicoke	М9Р
Kingsview Village, Martin Grove Gardens, Richvie	Etobicoke	M9R
Albion Gardens,Beaumond Heights,Humbergate,Jam	Etobicoke	M9V
Northwest	Etobicoke	M9W

103 rows × 1 columns

All the neighbourhoods are now concatenated under the postcode they are associated with. However, instead of regular index, the table has "Postcode" and "Borough" as its multi-column index. I am going to fix it below.

In [18]:

postcode_df.reset_index(inplace=True)
postcode_df

Out[18]:

Neighbourhood	Borough	Postcode	
Rouge,Malvern	Scarborough	M1B	0
Highland Creek,Rouge Hill,Port Union	Scarborough	M1C	1
Guildwood, Morningside, West Hill	Scarborough	M1E	2
Woburn	Scarborough	M1G	3
Cedarbrae	Scarborough	M1H	4
Weston	York	M9N	98
Westmount	Etobicoke	M9P	99
Kingsview Village, Martin Grove Gardens, Richvie	Etobicoke	M9R	100
Albion Gardens,Beaumond Heights,Humbergate,Jam	Etobicoke	M9V	101
Northwest	Etobicoke	M9W	102

103 rows × 3 columns

Get the size of the cleaned table.

In [19]:

print(postcode_df.shape)

(103, 3)

Next, find the latitude and longitude coordinates for each postcode.

In [20]:

Load the "Geospatial_Coordinates.csv" dataset
Geo_df=pd.read_csv("Geospatial_Coordinates.csv")
Geo_df.head()

Out[20]:

	Postcode	Latitude	Longitude
0	M1B	43.806686	-79.194353
1	M1C	43.784535	-79.160497
2	M1E	43.763573	-79.188711
3	M1G	43.770992	-79.216917
4	M1H	43.773136	-79.239476

In [21]:

Merge "postcode_df" and "Geo_df".

Use "Left" join so that all records from "postcode_df" and matched records from "Geo_df" are returned.

Geomerge_df=pd.merge(postcode_df,Geo_df, how="left")

Geomerge_df

Out[21]:

Postcode	Borough	Neighbourhood	Latitude	Longitude
M1B	Scarborough	Rouge,Malvern	43.806686	-79.194353
M1C	Scarborough	Highland Creek,Rouge Hill,Port Union	43.784535	-79.160497
M1E	Scarborough	Guildwood, Morningside, West Hill	43.763573	-79.188711
M1G	Scarborough	Woburn	43.770992	-79.216917
M1H	Scarborough	Cedarbrae	43.773136	-79.239476
M9N	York	Weston	43.706876	-79.518188
M9P	Etobicoke	Westmount	43.696319	-79.532242
M9R	Etobicoke	Kingsview Village, Martin Grove Gardens, Richvie	43.688905	-79.554724
M9V	Etobicoke	${\bf Albion\ Gardens, Beaumond\ Heights, Humbergate, Jam}$	43.739416	-79.588437
M9W	Etobicoke	Northwest	43.706748	-79.594054
	M1B M1C M1E M1G M1H M9N M9P M9R	M1B Scarborough M1C Scarborough M1E Scarborough M1G Scarborough M1H Scarborough M9N York M9P Etobicoke M9R Etobicoke M9V Etobicoke	M1B Scarborough Rouge,Malvern M1C Scarborough Highland Creek,Rouge Hill,Port Union M1E Scarborough Guildwood,Morningside,West Hill M1G Scarborough Woburn M1H Scarborough Cedarbrae M9N York Weston M9P Etobicoke Kingsview Village,Martin Grove Gardens,Richvie M9V Etobicoke Albion Gardens,Beaumond Heights,Humbergate,Jam	M1B Scarborough Rouge,Malvern 43.806686 M1C Scarborough Highland Creek,Rouge Hill,Port Union 43.784535 M1E Scarborough Guildwood,Morningside,West Hill 43.763573 M1G Scarborough Woburn 43.770992 M1H Scarborough Cedarbrae 43.773136 M9N York Weston 43.706876 M9P Etobicoke Kingsview Village,Martin Grove Gardens,Richvie 43.688905 M9V Etobicoke Albion Gardens,Beaumond Heights,Humbergate,Jam 43.739416

103 rows × 5 columns

Cluster the neighbourhoods and create visualizations.