```
In [1]: import pandas as pd
        import numpy as np
        from sklearn.cluster import KMeans
        import requests
In [2]: url="https://en.wikipedia.org/wiki/List of postal codes of Canada: M"
        header = {
          "User-Agent": "Mozilla/5.0 (X11; Linux x86 64) AppleWebKit/537.36 (KH
        TML, like Gecko) Chrome/50.0.2661.75 Safari/537.36",
          "X-Requested-With": "XMLHttpRequest"
        r = requests.get(url, headers=header)
        tables = pd.read html(r.text)
In [3]: df=pd.DataFrame(tables[0])
        # The dataframe will consist of three columns: PostalCode, Borough, and
         Neighborhood
        df.columns=['Postcode', 'Borough', 'Neighbourhood']
        df.drop([0],axis=0,inplace=True)
        df.reset index()
        # Only process the cells that have an assigned borough. Ignore cells wi
        th a borough that is Not assigned.
        df.drop(df[df['Borough']=="Not assigned"].index,axis=0, inplace=True)
        # More than one neighborhood can exist in one postal code area.
        # For example, in the table on the Wikipedia page,
        # you will notice that M5A is listed twice and has two neighborhoods:
```

```
# Harbourfront and Regent Park. These two rows will be combined into on
        e row with the neighborhoods
        # separated with a comma as shown in row 11 in the above table.
        df1=df.groupby("Postcode").agg(lambda x:','.join(set(x)))
        # If a cell has a borough but a Not assigned neighborhood,
        # then the neighborhood will be the same as the borough.
        # So for the 9th cell in the table on the Wikipedia page,
        # the value of the Borough and the Neighborhood columns will be Queen's
         Park.
        dfl.loc[dfl['Neighbourhood']=="Not assigned",'Neighbourhood']=dfl.loc[d
        f1['Neighbourhood']=="Not assigned",'Borough']
        df1.shape
Out[3]: (103, 2)
In [4]: from bs4 import BeautifulSoup
        soup = BeautifulSoup(requests.get(url).text,'lxml')
        My table = soup.find('table',{'class':'wikitable sortable'})
        My table
        table rows = My table.find all('tr')
        t=[]
        for tr in table rows:
            td = tr.find all('td')
            row = [tr.text.rstrip('\n') for tr in td]
            t.append(row)
        df=pd.DataFrame(t)
        df.columns=['Postcode', 'Borough', 'Neighbourhood']
        df.drop([0],axis=0,inplace=True)
        df.reset index()
```

```
# Only process the cells that have an assigned borough. Ignore cells wi
th a borough that is Not assigned.
df.drop(df[df['Borough']=="Not assigned"].index,axis=0, inplace=True)
# More than one neighborhood can exist in one postal code area.
# For example, in the table on the Wikipedia page,
# you will notice that M5A is listed twice and has two neighborhoods:
# Harbourfront and Regent Park. These two rows will be combined into on
e row with the neighborhoods
# separated with a comma as shown in row 11 in the above table.
df1=df.groupby("Postcode").agg(lambda x:','.join(set(x)))
# If a cell has a borough but a Not assigned neighborhood,
# then the neighborhood will be the same as the borough.
# So for the 9th cell in the table on the Wikipedia page,
# the value of the Borough and the Neighborhood columns will be Oueen's
Park.
dfl.loc[dfl['Neighbourhood']=="Not assigned",'Neighbourhood']=dfl.loc[d
f1['Neighbourhood']=="Not assigned",'Borough']
df1.shape
df1
```

Neighbourhood

## Out[4]:

	_0.009	
		Postcode
Malvern, Rouge	Scarborough	M1B
Rouge Hill, Port Union, Highland Creek	Scarborough	M1C
Guildwood, Morningside, West Hill	Scarborough	M1E
Woburn	Scarborough	M1G
Cedarbrae	Scarborough	M1H
Weston	York	M9N

Borough

## Postcode M9P Etobicoke Westmount M9R Etobicoke Kingsview Village, St. Phillips, Martin Grove ... M9V Etobicoke South Steeles, Silverstone, Humbergate, Jamest... M9W Etobicoke Northwest, West Humber - Clairville

103 rows × 2 columns

In [6]: geo\_data=pd.read\_csv("https://cocl.us/Geospatial\_data")
 geo\_data

## Out[6]:

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               98       M9N       43.706876       -79.518188         99       M9P       43.696319       -79.532242         100       M9R       43.688905       -79.554724         101       M9V       43.739416       -79.588437         102       M9W       43.706748       -79.594054			Postal Code	Latitude	Longitude
2       M1E       43.763573       -79.188711         3       M1G       43.770992       -79.216917         4       M1H       43.773136       -79.239476               98       M9N       43.706876       -79.518188         99       M9P       43.696319       -79.532242         100       M9R       43.688905       -79.554724         101       M9V       43.739416       -79.588437		0	M1B	43.806686	-79.194353
3 M1G 43.770992 -79.216917 4 M1H 43.773136 -79.239476 98 M9N 43.706876 -79.518188 99 M9P 43.696319 -79.532242 100 M9R 43.688905 -79.554724 101 M9V 43.739416 -79.588437		1	M1C	43.784535	-79.160497
4 M1H 43.773136 -79.239476  98 M9N 43.706876 -79.518188 99 M9P 43.696319 -79.532242 100 M9R 43.688905 -79.554724 101 M9V 43.739416 -79.588437		2	M1E	43.763573	-79.188711
<ul> <li>M9N 43.706876 -79.518188</li> <li>M9P 43.696319 -79.532242</li> <li>M9R 43.688905 -79.554724</li> <li>M9V 43.739416 -79.588437</li> </ul>		3	M1G	43.770992	-79.216917
98       M9N       43.706876       -79.518188         99       M9P       43.696319       -79.532242         100       M9R       43.688905       -79.554724         101       M9V       43.739416       -79.588437		4	M1H	43.773136	-79.239476
99       M9P       43.696319       -79.532242         100       M9R       43.688905       -79.554724         101       M9V       43.739416       -79.588437					
100       M9R       43.688905       -79.554724         101       M9V       43.739416       -79.588437		98	M9N	43.706876	-79.518188
<b>101</b> M9V 43.739416 -79.588437		99	М9Р	43.696319	-79.532242
		100	M9R	43.688905	-79.554724
<b>102</b> M9W 43.706748 -79.594054		101	M9V	43.739416	-79.588437
		102	M9W	43.706748	-79.594054

103 rows × 3 columns

```
In [7]: df1['Latitude']=geo_data['Latitude'].values
          df1['Longitude']=geo data['Longitude'].values
          df1
Out[7]:
                        Borough
                                                            Neighbourhood
                                                                             Latitude Longitude
           Postcode
                M1B Scarborough
                                                             Malvern, Rouge 43.806686 -79.194353
                                           Rouge Hill, Port Union, Highland Creek 43.784535 -79.160497
                M1C Scarborough
                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, St. Phillips, Martin Grove ... 43.688905 -79.554724
                M9V
                        Etobicoke South Steeles, Silverstone, Humbergate, Jamest... 43.739416 -79.588437
               M9W
                        Etobicoke
                                             Northwest, West Humber - Clairville 43.706748 -79.594054
          103 rows × 4 columns
In [8]:
          df1.shape
Out[8]: (103, 4)
In [ ]:
```