

BattleofNeighborhoods

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The Battle of the Neighborhoods

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2 Intoduction: Business Problem

Background Nowadays, there is a substantial increment of remote work due to the fact that the society is going through a pandemic. A lot of companies have offered the possibility to their employees to work from home (mainly). However, some employees could have had the idea of taking this opportunity to keep working for their employers but from other locations where in some way, could represent a better quality of life, or better personal or professional satisfaction.

People in Latin America that have enough level of incomes could have thought that this new way of working can be a opportunity to move to another country (where english is the official language) to learn a new language. A perfect combination has arised because those people could re locate to another country without the risk of losing their jobs.

Problem The question now is how a person that wants to learn english and is looking to move to a english spoken country can find the best option? Over the internet, there are a lot of courses offered in different cities, so, the main purpose of this exercise is finding the best one taking into account the variety of venues around the course location as a decisor.

It is important to aclare that this exercise assumes that the closer the housing to the course location the better for the “new student”.

3 Data sources

Where do we find the data?

The two main data sources are:

- Web scraping
- Foursquare API

The exercise is made for people who is looking to move to Canada or United States so, the website that is used to scrape the data of the courses will only be for USA and Canada.

Only the best 15 courses for each country according to the website will be considered

The URL for the courses in **Canada**: <https://www.languageinternational.com/english-courses-canada>

The URL for the courses in **United States**: <https://www.languageinternational.com/english-courses-usa>

If we look at the web sites, the main page does not have the addresses of the institutes, so, the first task is finding their addresses within the website.

How it will be used?

Once all the addresses are saved in the dataframe of the courses (the structure of the dataframe will be **institutenam**e, **city**, **name**, **address**, **postalcode**), the next step is finding the coordinates of each institute.

After having all the neccesary data of the courses with their specific addresses the next step is use Foursquare as the provider of the venues around specific locations (through the API)

Note: the locations that are going to be used in the API are the institue ones.

4 Data extraction

Let's import necessary Libraries

```
[2]: # This section is used to import libraries that are going to be used along the
    ↪project

!pip install bs4
from bs4 import BeautifulSoup # this module helps in web scrapping.

!conda install -c conda-forge geopy --yes
from geopy.geocoders import Nominatim # convert an address into latitude and
    ↪longitude values

!conda install -c conda-forge folium=0.5.0 --yes
```

```

import folium # map rendering library

import numpy as np # library to handle data in a vectorized manner

import pandas as pd # library for data analysis

import re # library for regular expressions

import json # library to handle JSON files

import requests # library to handle requests
from pandas.io.json import json_normalize # tranform JSON file into a pandas
↳dataframe

# Matplotlib and associated plotting modules
import matplotlib.cm as cm
import matplotlib.colors as colors
import matplotlib.pyplot as plt

#!conda install -c conda-forge plotly
#import plotly.express as px
#import plotly.graph_objects as go

# import k-means from clustering stage
from sklearn.cluster import KMeans

from IPython.display import Image

print('Libraries imported.')

```

```

Requirement already satisfied: bs4 in
c:\users\sebastian.bedoya\anaconda3\lib\site-packages (0.0.1)
Requirement already satisfied: beautifulsoup4 in
c:\users\sebastian.bedoya\anaconda3\lib\site-packages (from bs4) (4.9.3)
Requirement already satisfied: soupsieve>1.2; python_version >= "3.0" in
c:\users\sebastian.bedoya\anaconda3\lib\site-packages (from beautifulsoup4->bs4)
(2.0.1)
Collecting package metadata (current_repodata.json): ...working... done
Solving environment: ...working... done

```

```
# All requested packages already installed.
```

```

Collecting package metadata (current_repodata.json): ...working... done
Solving environment: ...working... done

```

```
# All requested packages already installed.
```

```
Libraries imported.
```

Now it's time to start doing the web scraping to obtain the addresses of the institutes

The first url to scrape is for the courses of Canada

```
[3]: urlcanada="https://www.languageinternational.com/english-courses-canada"

html_data_canada = requests.get(urlcanada).text
soup_canada = BeautifulSoup(html_data_canada,"html5lib")
```

Let's create a dataframe where the scraped data is going to be stored.

The dataframe will have 6 columns

```
[4]: df_courses = pd.DataFrame(columns=["Location", "CourseName", "InstituteName",
    ↳ "PriceFrom", "PriceTo", "Address", "urlcourse", "urlinstitute", "latitude",
    ↳ "longitude", "Country"])
```

All the data that we need is inside the div section with the class: **row gutterh-xs**

```
[5]: for div in soup_canada.find_all('div',class_='row gutterh-xs'):

    location = div.find_all(attrs={'class': 'searchresult_subh'})[0].text
    course_name = div.find_all(itemprop='name')[0].text
    institutename = div.find_all(itemprop='name')[1].text
    pricefrom = div.find_all(attrs={'class':
    ↳ 'courselist_course_prices_box_amount'})[0].text
    priceto = div.find_all(attrs={'class':
    ↳ 'courselist_course_prices_box_amount'})[1].text

    url = div.find_all('a', href=True)[0]

    df_courses = df_courses.append({"Location":location, "CourseName":
    ↳ course_name, "InstituteName":institutename, "PriceFrom":pricefrom, "PriceTo":
    ↳ priceto, "urlcourse":url['href'], "Country": "Canada"}, ignore_index=True)
```

The columns that contains prices need to be cleaned.

- Regex patterns are used to replace some strings present in both columns

```
[6]: df_courses[['PriceFrom','PriceTo']] = df_courses[['PriceFrom','PriceTo']].
    ↳ replace(to_replace=["\\t|\\n|\\r", "\\W"], value=["", ""], regex=True)
```

Here the result of the web scraping from the url for the courses in Canada

```
[7]: df_courses.head()
```

```
[7]:
```

	Location	CourseName	\
0	Calgary, Canada	General English 20 (GE20)	
1	Hamilton, Canada	ESL Private Lessons	
2	Calgary, Canada	General English (GE25)	
3	Hamilton, Canada	Intensive English	

4 Vancouver, Canada

General 20

	InstituteName	PriceFrom	PriceTo	Address	\
0	ANNE'S Language House	354	687	NaN	
1	Metropolitan College	384	619	NaN	
2	ANNE'S Language House	384	717	NaN	
3	Metropolitan College	290	525	NaN	
4	Language Studies International (LSI): Vancouver	431	759	NaN	

	urlcourse	urlinstitute	latitude	\
0	/course/general-english-20-ge20-anne-s-languag...	NaN	NaN	
1	/course/esl-private-lessons-metropolitan-colle...	NaN	NaN	
2	/course/general-english-ge25-anne-s-language-h...	NaN	NaN	
3	/course/intensive-english-metropolitan-college...	NaN	NaN	
4	/course/general-20-language-studies-internatio...	NaN	NaN	

	longitude	Country
0	NaN	Canada
1	NaN	Canada
2	NaN	Canada
3	NaN	Canada
4	NaN	Canada

Now, let's do exactly the same but for the courses that are offered in the United States

```
[8]: urlusa="https://www.languageinternational.com/english-courses-usa"
```

```
html_data_usa = requests.get(urlusa).text
soup_usa = BeautifulSoup(html_data_usa,"html5lib")
```

```
[9]: for div in soup_usa.find_all('div',class_='row gutterh-xs'):

    location = div.find_all(attrs={'class': 'searchresult_subh'})[0].text
    coursename = div.find_all(itemprop='name')[0].text
    institutename = div.find_all(itemprop='name')[1].text
    pricefrom = div.find_all(attrs={'class': '
↪'courselist_course_prices_box_amount'})[0].text
    priceto = div.find_all(attrs={'class': '
↪'courselist_course_prices_box_amount'})[1].text

    url = div.find_all('a', href=True)[0]

    df_courses = df_courses.append({"Location":location, "CourseName":
↪coursename, "InstituteName":institutename, "PriceFrom":pricefrom, "PriceTo":
↪priceto, "urlcourse":url['href'], "Country":"USA"}, ignore_index=True)
```

```
[10]: df_courses[['PriceFrom', 'PriceTo']] = df_courses[['PriceFrom', 'PriceTo']].
      ↪replace(to_replace=["\t|\n|\r", "\W"], value=["", ""], regex=True)
```

Now we have the dataframe with the data of Canada and United States courses

```
[11]: df_courses
```

```
[11]:
```

	Location	CourseName \
0	Calgary, Canada	General English 20 (GE20)
1	Hamilton, Canada	ESL Private Lessons
2	Calgary, Canada	General English (GE25)
3	Hamilton, Canada	Intensive English
4	Vancouver, Canada	General 20
5	Toronto, Canada	Intensive 30
6	Vancouver, Canada	Intensive 30
7	Toronto, Canada	General 20
8	Vancouver, Canada	Intensive 25
9	Toronto, Canada	Club 40+ (20 lessons per week plus afternoon a...
10	Toronto, Canada	Afternoon 10
11	Vancouver, Canada	Club 40+ (20 lessons per week plus afternoon a...
12	Vancouver, Canada	Afternoon 10
13	Vancouver, Canada	General English - Intensive
14	Montreal, Canada	Super Intensive English
15	San Diego, USA	English Max Course (18 hrs/week)
16	San Diego, USA	English Focus Course (12 hrs/week)
17	San Diego, USA	English Max
18	San Diego, USA	English Max Course (18 hrs/week)
19	San Diego, USA	English Focus Course (12 hrs/week)
20	Los Angeles, USA	ESL Program (12 Weeks Minimum)
21	Boston, USA	Intensive 30
22	Boston, USA	Club 40+ (20 lessons per week plus afternoon a...
23	San Diego, USA	Intensive 30
24	New York City, USA	One-to-One (20 lessons per week)
25	Boston, USA	One-to-One (5 lessons per week)
26	Boston, USA	General 20
27	New York City, USA	One-to-One (10 lessons per week)
28	San Diego, USA	Club 40+ (20 lessons per week plus afternoon a...
29	San Diego, USA	General 20

	InstituteName	PriceFrom	PriceTo \
0	ANNE'S Language House	354	687
1	Metropolitan College	384	619
2	ANNE'S Language House	384	717
3	Metropolitan College	290	525
4	Language Studies International (LSI): Vancouver	431	759
5	Language Studies International (LSI): Toronto	495	823
6	Language Studies International (LSI): Vancouver	495	823
7	Language Studies International (LSI): Toronto	431	759

8	Language Studies International (LSI): Vancouver	465	794
9	Language Studies International (LSI): Toronto	708	1037
10	Language Studies International (LSI): Toronto	286	614
11	Language Studies International (LSI): Vancouver	708	1037
12	Language Studies International (LSI): Vancouver	286	614
13	International Language Academy of Canada Vanco...	503	870
14	Bouchereau Lingua International	533	887
15	Connect English Language Institute- Mission Va...	390	None
16	Connect English- La Jolla	365	None
17	Connect English Language Institute- San Diego ...	390	None
18	Connect English- La Jolla	390	None
19	Connect English Language Institute- Mission Va...	315	None
20	American English Language School	335	None
21	Language Studies International (LSI): Boston	645	960
22	Language Studies International (LSI): Boston	840	1155
23	Language Studies International (LSI): San Diego	645	950
24	Language Studies International (LSI): New York	1955	2280
25	Language Studies International (LSI): Boston	575	890
26	Language Studies International (LSI): Boston	535	850
27	Language Studies International (LSI): New York	1055	1380
28	Language Studies International (LSI): San Diego	840	1240
29	Language Studies International (LSI): San Diego	535	935

	Address	urlcourse	urlinstitute \
0	NaN /course/general-english-20-ge20-anne-s-languag...		NaN
1	NaN /course/esl-private-lessons-metropolitan-colle...		NaN
2	NaN /course/general-english-ge25-anne-s-language-h...		NaN
3	NaN /course/intensive-english-metropolitan-college...		NaN
4	NaN /course/general-20-language-studies-internatio...		NaN
5	NaN /course/intensive-30-language-studies-internat...		NaN
6	NaN /course/intensive-30-language-studies-internat...		NaN
7	NaN /course/general-20-language-studies-internatio...		NaN
8	NaN /course/intensive-25-language-studies-internat...		NaN
9	NaN /course/club-40-20-lessons-per-week-plus-after...		NaN
10	NaN /course/afternoon-10-language-studies-internat...		NaN
11	NaN /course/club-40-20-lessons-per-week-plus-after...		NaN
12	NaN /course/afternoon-10-language-studies-internat...		NaN
13	NaN /course/general-english-intensive-international...		NaN
14	NaN /course/super-intensive-english-bouchereau-lin...		NaN
15	NaN /course/english-max-course-18-hrs-week-connect...		NaN
16	NaN /course/english-focus-course-12-hrs-week-conne...		NaN
17	NaN /course/english-max-connect-english-language-i...		NaN
18	NaN /course/english-max-course-18-hrs-week-connect...		NaN
19	NaN /course/english-focus-course-12-hrs-week-conne...		NaN
20	NaN /course/esl-program-12-weeks-minimum-american-...		NaN
21	NaN /course/intensive-30-language-studies-internat...		NaN
22	NaN /course/club-40-20-lessons-per-week-plus-after...		NaN

23	NaN	/course/intensive-30-language-studies-internat...	NaN
24	NaN	/course/one-to-one-20-lessons-per-week-languag...	NaN
25	NaN	/course/one-to-one-5-lessons-per-week-language...	NaN
26	NaN	/course/general-20-language-studies-internatio...	NaN
27	NaN	/course/one-to-one-10-lessons-per-week-languag...	NaN
28	NaN	/course/club-40-20-lessons-per-week-plus-after...	NaN
29	NaN	/course/general-20-language-studies-internatio...	NaN

	latitude	longitude	Country
0	NaN	NaN	Canada
1	NaN	NaN	Canada
2	NaN	NaN	Canada
3	NaN	NaN	Canada
4	NaN	NaN	Canada
5	NaN	NaN	Canada
6	NaN	NaN	Canada
7	NaN	NaN	Canada
8	NaN	NaN	Canada
9	NaN	NaN	Canada
10	NaN	NaN	Canada
11	NaN	NaN	Canada
12	NaN	NaN	Canada
13	NaN	NaN	Canada
14	NaN	NaN	Canada
15	NaN	NaN	USA
16	NaN	NaN	USA
17	NaN	NaN	USA
18	NaN	NaN	USA
19	NaN	NaN	USA
20	NaN	NaN	USA
21	NaN	NaN	USA
22	NaN	NaN	USA
23	NaN	NaN	USA
24	NaN	NaN	USA
25	NaN	NaN	USA
26	NaN	NaN	USA
27	NaN	NaN	USA
28	NaN	NaN	USA
29	NaN	NaN	USA

The column **Address** and **urlInstitute** as you can see, have **NaN** values because the address and the url of the institute is not present in the scraped url.

The address can be found once the course is selected. A new page needs to be scraped for each one of the courses present in the dataframe

The url format to scrape data related to the course for all the rows is <https://www.languageinternational.com/xxxxxxxxxx>

Where *xxxxxxxxxx* will be replaced with the **urlcourse column** to scrape the address and url of the institute of each course

Note: the url of the institute is scraped because the data of the latitude and longitude is already present in the web site in the page of the institutes

```
[12]: for index, row in df_courses.iterrows():

    urlcourse="https://www.languageinternational.com" + row['urlcourse']
    html_data_course = requests.get(urlcourse).text
    soup_course = BeautifulSoup(html_data_course,"html5lib")

    for div in soup_course.find_all(attrs={'class': 'school_upper'}):
        df_courses.iloc[index]['Address'] = div.find_all('span')[0].text
        print('Address: {} '.format(div.find_all('span')[0].text))

        for div in soup_course.find_all(attrs={'class': 'pageheader_text'}):
            df_courses.iloc[index]['urlinstitute'] = div.find_all('a',
↪href=True)[0]['href']
            print('url of institue: {} '.format(div.find_all('a',
↪href=True)[0]['href']))
            print('--- Added to dataframe ---')

print('-----')
print('--End of update--')
```

```
Address: 101 6th Avenue S.W., Suite 1250, Calgary, Alberta, AB T2P3P4, Canada
url of institue: /school/anne-s-language-house-64694
--- Added to dataframe ---
Address: 146 James Street South, Hamilton, Ontario,, Hamilton, Ontario L8P 3A2,
Canada
url of institue: /school/metropolitan-college-64750
--- Added to dataframe ---
Address: 101 6th Avenue S.W., Suite 1250, Calgary, Alberta, AB T2P3P4, Canada
url of institue: /school/anne-s-language-house-64694
--- Added to dataframe ---
Address: 146 James Street South, Hamilton, Ontario,, Hamilton, Ontario L8P 3A2,
Canada
url of institue: /school/metropolitan-college-64750
--- Added to dataframe ---
Address: 101-808 Nelson Street, Vancouver, BC V6Z 2H2, Canada
url of institue: /school/language-studies-international-lsi-vancouver-18606
--- Added to dataframe ---
Address: 1055 Yonge Street, Suite #210, Toronto, ON M4W 2L2, Canada
url of institue: /school/language-studies-international-lsi-toronto-18605
--- Added to dataframe ---
Address: 101-808 Nelson Street, Vancouver, BC V6Z 2H2, Canada
url of institue: /school/language-studies-international-lsi-vancouver-18606
```

```

--- Added to dataframe ---
Address: 1055 Yonge Street, Suite #210, Toronto, ON M4W 2L2, Canada
url of institue: /school/language-studies-international-lsi-toronto-18605
--- Added to dataframe ---
Address: 101-808 Nelson Street, Vancouver, BC V6Z 2H2, Canada
url of institue: /school/language-studies-international-lsi-vancouver-18606
--- Added to dataframe ---
Address: 1055 Yonge Street, Suite #210, Toronto, ON M4W 2L2, Canada
url of institue: /school/language-studies-international-lsi-toronto-18605
--- Added to dataframe ---
Address: 1055 Yonge Street, Suite #210, Toronto, ON M4W 2L2, Canada
url of institue: /school/language-studies-international-lsi-toronto-18605
--- Added to dataframe ---
Address: 101-808 Nelson Street, Vancouver, BC V6Z 2H2, Canada
url of institue: /school/language-studies-international-lsi-vancouver-18606
--- Added to dataframe ---
Address: 101-808 Nelson Street, Vancouver, BC V6Z 2H2, Canada
url of institue: /school/language-studies-international-lsi-vancouver-18606
--- Added to dataframe ---
Address: 1199 West Pender Street, Vancouver, British Columbia V6E 2R1, Canada
url of institue: /school/international-language-academy-of-canada-
vancouver-63381
--- Added to dataframe ---
Address: 70 Notre Dame West, Suite 400, Montréal, Quebec H2Y 1S6, Canada
url of institue: /school/bouchereau-lingua-international-63865
--- Added to dataframe ---
Address: 4560 Alvarado Canyon Road, Suite 2B, San Diego, California 92120, USA
url of institue: /school/connect-english-language-institute-mission-valley-63248
--- Added to dataframe ---
Address: 5090 Shoreham Place, Suite 206, San Diego, California 92037, USA
url of institue: /school/connect-english-la-jolla-64046
--- Added to dataframe ---
Address: 3565 Del Rey St., Suite 300, San Diego, California 92109, USA
url of institue: /school/connect-english-language-institute-san-diego-pacific-
beach-campus-64605
--- Added to dataframe ---
Address: 5090 Shoreham Place, Suite 206, San Diego, California 92037, USA
url of institue: /school/connect-english-la-jolla-64046
--- Added to dataframe ---
Address: 4560 Alvarado Canyon Road, Suite 2B, San Diego, California 92120, USA
url of institue: /school/connect-english-language-institute-mission-valley-63248
--- Added to dataframe ---
Address: 3230 E. Imperial Hwy, Suite 301, Brea, California 92821, USA
url of institue: /school/american-english-language-school-68322
--- Added to dataframe ---
Address: 105 Beach Street, Boston, MA 02111, USA
url of institue: /school/language-studies-international-lsi-boston-68
--- Added to dataframe ---

```

```

Address: 105 Beach Street, Boston, MA 02111, USA
url of institue: /school/language-studies-international-lsi-boston-68
--- Added to dataframe ---
Address: 1706 5th Avenue, San Diego, CA 92101, USA
url of institue: /school/language-studies-international-lsi-san-diego-231
--- Added to dataframe ---
Address: 40 Rector Street, 10th Floor, Suite 1000, New York, NY 10006, USA
url of institue: /school/language-studies-international-lsi-new-york-232
--- Added to dataframe ---
Address: 105 Beach Street, Boston, MA 02111, USA
url of institue: /school/language-studies-international-lsi-boston-68
--- Added to dataframe ---
Address: 105 Beach Street, Boston, MA 02111, USA
url of institue: /school/language-studies-international-lsi-boston-68
--- Added to dataframe ---
Address: 40 Rector Street, 10th Floor, Suite 1000, New York, NY 10006, USA
url of institue: /school/language-studies-international-lsi-new-york-232
--- Added to dataframe ---
Address: 1706 5th Avenue, San Diego, CA 92101, USA
url of institue: /school/language-studies-international-lsi-san-diego-231
--- Added to dataframe ---
Address: 1706 5th Avenue, San Diego, CA 92101, USA
url of institue: /school/language-studies-international-lsi-san-diego-231
--- Added to dataframe ---
-----
--End of update-

```

```
[13]: df_courses.head()
```

```

[13]:
      Location      CourseName \
0  Calgary, Canada  General English 20 (GE20)
1  Hamilton, Canada      ESL Private Lessons
2  Calgary, Canada      General English (GE25)
3  Hamilton, Canada      Intensive English
4  Vancouver, Canada      General 20

      InstituteName  PriceFrom  PriceTo \
0      ANNE'S Language House      354      687
1      Metropolitan College      384      619
2      ANNE'S Language House      384      717
3      Metropolitan College      290      525
4  Language Studies International (LSI): Vancouver      431      759

      Address \
0  101 6th Avenue S.W., Suite 1250, Calgary, Albe...
1  146 James Street South, Hamilton, Ontario,, Ha...
2  101 6th Avenue S.W., Suite 1250, Calgary, Albe...

```

```

3 146 James Street South, Hamilton, Ontario,, Ha...
4 101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...

```

```

                                urlcourse \
0 /course/general-english-20-ge20-anne-s-languag...
1 /course/esl-private-lessons-metropolitan-colle...
2 /course/general-english-ge25-anne-s-language-h...
3 /course/intensive-english-metropolitan-college...
4 /course/general-20-language-studies-internatio...

```

```

                                urlinstitute latitude longitude \
0 /school/anne-s-language-house-64694 NaN NaN
1 /school/metropolitan-college-64750 NaN NaN
2 /school/anne-s-language-house-64694 NaN NaN
3 /school/metropolitan-college-64750 NaN NaN
4 /school/language-studies-international-lsi-van... NaN NaN

```

```

Country
0 Canada
1 Canada
2 Canada
3 Canada
4 Canada

```

At this point, in the dataframe, the columns *address* and *urlinstitute* are already filled up with the scraped data

With this data ready, we can proceed to get the coordinates from the website as well. The coordinates will be added to the dataframe

Note: in this step, a regular expression is used to extract a portion of the entire text where the coordinates are

```

[14]: for index, row in df_courses.iterrows():
        urlinstitute="https://www.languageinternational.com" + row['urlinstitute']
        html_data_institute = requests.get(urlinstitute).text

        # Create a pattern to match names
        pattern = re.compile(r'(?=latitude)(.)*(?=url)', flags = re.M)
        # Find all occurrences of the pattern
        result = pattern.findall(html_data_institute)
        # Let's clean the text matched
        mapping = {'latitude':'', 'longitude':'', '':'', ':':''}
        for k, v in mapping.items():
            result[0] = result[0].replace(k, v)

        # Now, a split is used to assign latitude and longitude variables
        latitude = result[0].split(',')[0]

```

```

df_courses.iloc[index]['latitude'] = result[0].split(',')[0]

longitude = result[0].split(',')[1]
df_courses.iloc[index]['longitude'] = result[0].split(',')[1]

print(latitude, longitude)
print('--- Lat and Lon added ---')

print('-----')
print('--End of update--')

```

```

51.04715 -114.06334
--- Lat and Lon added ---
43.25243 -79.8713
--- Lat and Lon added ---
51.04715 -114.06334
--- Lat and Lon added ---
43.25243 -79.8713
--- Lat and Lon added ---
49.28004 -123.12491
--- Lat and Lon added ---
43.67851 -79.38973
--- Lat and Lon added ---
49.28004 -123.12491
--- Lat and Lon added ---
43.67851 -79.38973
--- Lat and Lon added ---
49.28004 -123.12491
--- Lat and Lon added ---
43.67851 -79.38973
--- Lat and Lon added ---
43.67851 -79.38973
--- Lat and Lon added ---
49.28004 -123.12491
--- Lat and Lon added ---
49.28004 -123.12491
--- Lat and Lon added ---
49.28836 -123.12259
--- Lat and Lon added ---
45.466221 -74.077549
--- Lat and Lon added ---
32.78098 -117.09628
--- Lat and Lon added ---
32.85252 -117.18622
--- Lat and Lon added ---
32.80372 -117.21411
--- Lat and Lon added ---
32.85252 -117.18622

```

```

--- Lat and Lon added ---
32.78098 -117.09628
--- Lat and Lon added ---
33.90986 -117.85417
--- Lat and Lon added ---
42.35072 -71.05826
--- Lat and Lon added ---
42.35072 -71.05826
--- Lat and Lon added ---
32.72327 -117.16051
--- Lat and Lon added ---
40.70863 -74.01466
--- Lat and Lon added ---
42.35072 -71.05826
--- Lat and Lon added ---
42.35072 -71.05826
--- Lat and Lon added ---
40.70863 -74.01466
--- Lat and Lon added ---
32.72327 -117.16051
--- Lat and Lon added ---
32.72327 -117.16051
--- Lat and Lon added ---
-----
--End of update-

```

The dataframe now is complete!

```
[15]: df_courses.head()
```

```

[15]:
      Location      CourseName \
0    Calgary, Canada  General English 20 (GE20)
1    Hamilton, Canada    ESL Private Lessons
2    Calgary, Canada    General English (GE25)
3    Hamilton, Canada    Intensive English
4    Vancouver, Canada    General 20

      InstituteName  PriceFrom  PriceTo \
0      ANNE'S Language House      354      687
1      Metropolitan College      384      619
2      ANNE'S Language House      384      717
3      Metropolitan College      290      525
4  Language Studies International (LSI): Vancouver      431      759

      Address \
0  101 6th Avenue S.W., Suite 1250, Calgary, Albe...
1  146 James Street South, Hamilton, Ontario,, Ha...
2  101 6th Avenue S.W., Suite 1250, Calgary, Albe...

```

```

3 146 James Street South, Hamilton, Ontario,, Ha...
4 101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...

```

```

                                urlcourse \
0 /course/general-english-20-ge20-anne-s-languag...
1 /course/esl-private-lessons-metropolitan-colle...
2 /course/general-english-ge25-anne-s-language-h...
3 /course/intensive-english-metropolitan-college...
4 /course/general-20-language-studies-internatio...

```

```

                                urlinstitute  latitude  longitude \
0 /school/anne-s-language-house-64694  51.04715  -114.06334
1 /school/metropolitan-college-64750  43.25243   -79.8713
2 /school/anne-s-language-house-64694  51.04715  -114.06334
3 /school/metropolitan-college-64750  43.25243   -79.8713
4 /school/language-studies-international-lsi-van... 49.28004  -123.12491

```

```

Country
0 Canada
1 Canada
2 Canada
3 Canada
4 Canada

```

5 Institutes visualization and Data

Looking at the data, it is suitable to have a child dataframe with the institutes with their respective coordinates to avoid duplicates. This happens because it can be that a institute can offer more than one course.

```

[16]: df_institutes = df_courses[['InstituteName','latitude','longitude', 'Country']].
      ↪drop_duplicates(subset=['InstituteName','latitude','longitude', 'Country'],
      ↪keep='last').reset_index()
df_institutes

```

```

[16]:   index      InstituteName  latitude \
0      2  ANNE'S Language House  51.04715
1      3  Metropolitan College  43.25243
2     10  Language Studies International (LSI): Toronto 43.67851
3     12  Language Studies International (LSI): Vancouver 49.28004
4     13  International Language Academy of Canada Vanco... 49.28836
5     14  Bouchereau Lingua International 45.466221
6     17  Connect English Language Institute- San Diego ... 32.80372
7     18  Connect English- La Jolla 32.85252
8     19  Connect English Language Institute- Mission Va... 32.78098
9     20  American English Language School 33.90986
10    26  Language Studies International (LSI): Boston 42.35072

```

```

11      27      Language Studies International (LSI): New York    40.70863
12      29      Language Studies International (LSI): San Diego    32.72327

```

```

      longitude Country
0  -114.06334  Canada
1   -79.8713  Canada
2   -79.38973  Canada
3  -123.12491  Canada
4  -123.12259  Canada
5   -74.077549  Canada
6  -117.21411   USA
7  -117.18622   USA
8  -117.09628   USA
9  -117.85417   USA
10 -71.05826   USA
11 -74.01466   USA
12 -117.16051   USA

```

```

[17]: df_institutes["latitude"] = df_institutes["latitude"].astype("float")
      df_institutes["longitude"] = df_institutes["longitude"].astype("float")

      print(df_institutes.dtypes)

```

```

index          int64
InstituteName   object
latitude        float64
longitude        float64
Country          object
dtype: object

```

```

[18]: print('--> We can observe now that there are ', df_institutes.shape[0], '
      ↪different institutes that offer the best 30 courses in Canada and United
      ↪States <--')

```

```

--> We can observe now that there are 13 different institutes that offer the
best 30 courses in Canada and United States <--

```

```

[19]: bar_data = df_institutes.groupby(['Country'])['InstituteName'].count().
      ↪reset_index()
      bar_data

```

```

[19]: Country  InstituteName
0  Canada          6
1   USA           7

```

```

[20]: fig, ax = plt.subplots(figsize=(5, 3))

      ax.bar(bar_data['Country'], bar_data['InstituteName'])

```

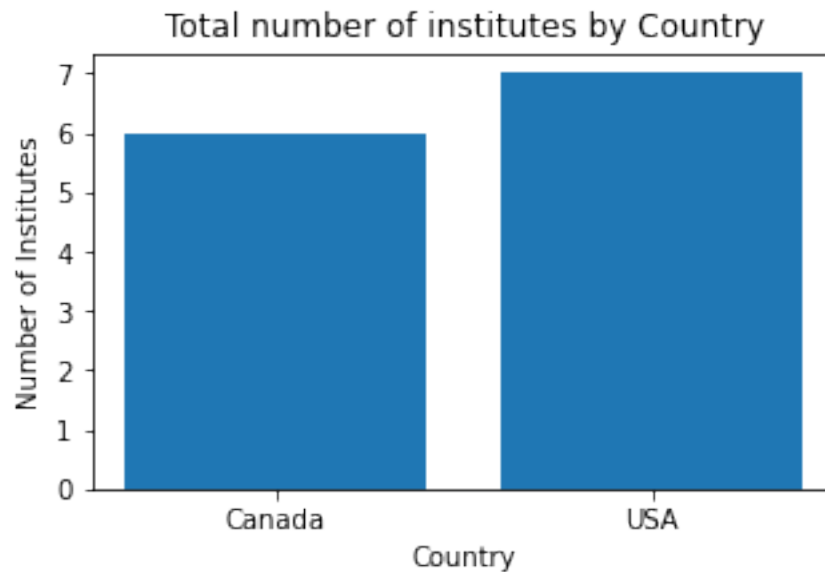


```

ax.set_xlabel('Country')
ax.set_ylabel('Number of Institutes')

plt.title('Total number of institutes by Country')
plt.show()
print('Done!')

```



Done!

Let's get the geographical coordinates of North America.

```

[21]: address = 'North America'

geolocator = Nominatim(user_agent="northamerica_explorer")
location = geolocator.geocode(address)
latitude = location.latitude
longitude = location.longitude

print('The geograpical coordinate of North America are {}, {}.'.
      ↪format(latitude, longitude))

```

The geograpical coordinate of North America are 51.0000002, -109.0.

Create a map of North America with the institutes superimposed on top

```

[23]: courses_map = folium.Map(location=[latitude, longitude], zoom_start=4)

```

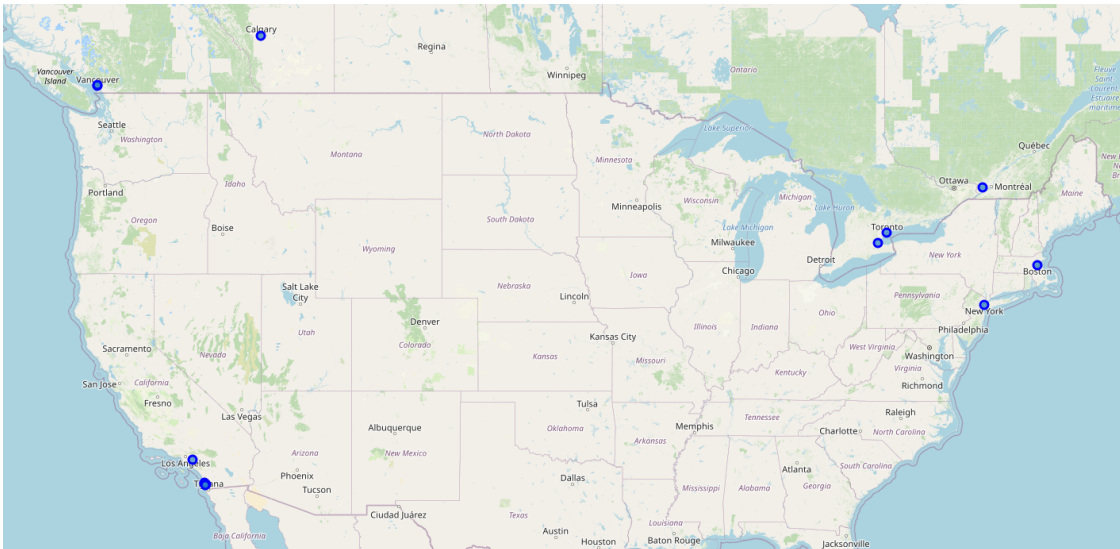
```

for lat, lng, institute, country in zip(df_institutes['latitude'],
↳df_institutes['longitude'], df_institutes['InstituteName'],
↳df_institutes['Country']):
    label = institute
    label = folium.Popup(label, parse_html=True)
    folium.CircleMarker(
        [lat, lng],
        radius=5,
        popup=label,
        color='blue',
        fill=True,
        fill_color='#3186cc',
        fill_opacity=0.7,
        parse_html=False).add_to(courses_map)

# display map
courses_map
#Image(filename='mapInstitutes.png')

```

[23]:



Define Foursquare Credentials and Version

Next, we are going to start utilizing the Foursquare API to explore the neighborhoods and segment them.

```

[24]: CLIENT_ID = 'IJFZ0DC4A2GVGVYMK0QCF413Q114Y1GSJX3KPOT44C4C3JGI' # your
↳Foursquare ID
CLIENT_SECRET = 'E50AREFRWDXT1FY3SAYQJZEAGNX4S4VUTIO1N4WNELI4HF3J' # your
↳Foursquare Secret
VERSION = '20180605' # Foursquare API version

```

```
LIMIT = 100 # A default Foursquare API limit value
```

```
print('Your credentails:')  
print('CLIENT_ID: ' + CLIENT_ID)  
print('CLIENT_SECRET:' + CLIENT_SECRET)
```

Your credentails:

CLIENT_ID: IJFZ0DC4A2GVGVYMKOQCF413Q114Y1GSJX3KPOT44C4C3JGI

CLIENT_SECRET:E50AREFRWDXT1FY3SAYQJZEAGNX4S4VUTIO1N4WNELI4HF3J

Let's create a function to get the data from the venues of the entire list of institutes

```
[25]: 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={}'.format(  
            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 = ['InstituteName',  
                             'InstituteLatitude',  
                             'InstituteLongitude',  
                             'Venue',
```

```

        'VenueLatitude',
        'VenueLongitude',
        'VenueCategory']

    return(nearby_venues)

```

Let's run the above function on each institute and create a new dataframe called `institutes_venues`

```

[26]: # type your answer here
institutes_venues = getNearbyVenues(names=df_institutes['InstituteName'],
                                   latitudes=df_institutes['latitude'],
                                   longitudes=df_institutes['longitude']
                                   )

```

```

ANNE'S Language House
Metropolitan College
Language Studies International (LSI): Toronto
Language Studies International (LSI): Vancouver
International Language Academy of Canada Vancouver
Bouchereau Lingua International
Connect English Language Institute- San Diego (Pacific Beach Campus)
Connect English- La Jolla
Connect English Language Institute- Mission Valley
American English Language School
Language Studies International (LSI): Boston
Language Studies International (LSI): New York
Language Studies International (LSI): San Diego

```

Let's check the size of the resulting dataframe

```

[27]: print(institutes_venues.shape)
      institutes_venues.head()

```

```

(515, 7)

```

```

[27]:
      InstituteName  InstituteLatitude  InstituteLongitude  \
0  ANNE'S Language House           51.04715          -114.06334
1  ANNE'S Language House           51.04715          -114.06334
2  ANNE'S Language House           51.04715          -114.06334
3  ANNE'S Language House           51.04715          -114.06334
4  ANNE'S Language House           51.04715          -114.06334

      Venue  VenueLatitude  VenueLongitude  \
0  The Palomino Smokehouse    51.046435    -114.063410
1                Blink      51.045422    -114.063733
2  Phil & Sebastian Coffee Roasters    51.045619    -114.063324
3                Over Easy Breakfast    51.048561    -114.065917
4      Hyatt Regency Calgary    51.046373    -114.062583

```

```

VenueCategory
0  American Restaurant
1      Restaurant
2      Coffee Shop
3  Breakfast Spot
4      Hotel

```

Let's check how many venues were returned for each institute

```
[28]: institutes_venues.groupby('InstituteName')['InstituteName'].count()
```

```
[28]: InstituteName
ANNE'S Language House                                41
American English Language School                     21
Bouchereau Lingua International                       5
Connect English Language Institute- Mission Valley   16
Connect English Language Institute- San Diego (Pacific Beach Campus) 13
Connect English- La Jolla                             6
International Language Academy of Canada Vancouver   45
Language Studies International (LSI): Boston          82
Language Studies International (LSI): New York        100
Language Studies International (LSI): San Diego        23
Language Studies International (LSI): Toronto          42
Language Studies International (LSI): Vancouver        89
Metropolitan College                                32
Name: InstituteName, dtype: int64
```

Let's find out how many unique categories can be curated from all the returned venues

```
[33]: print('There are {} uniques categories.'.
        ↪format(len(institutes_venues['VenueCategory'].unique())))
```

There are 151 uniques categories.

6 Preproccesing steps: one hot encoding

```
[34]: institutes_onehot = pd.get_dummies(institutes_venues[['VenueCategory']],
        ↪prefix="", prefix_sep="")

# add institute column back to dataframe
institutes_onehot['InstituteName'] = institutes_venues['InstituteName']

# move InstituteName column to the first column
fixed_columns = [institutes_onehot.columns[-1]] + list(institutes_onehot.
        ↪columns[:-1])
institutes_onehot = institutes_onehot[fixed_columns]
```

```
institutes_onehot.head()
```

```
[34]:
```

	InstituteName	Accessories	Store	American Restaurant	Art Gallery	\
0	ANNE'S Language House		0	1	0	
1	ANNE'S Language House		0	0	0	
2	ANNE'S Language House		0	0	0	
3	ANNE'S Language House		0	0	0	
4	ANNE'S Language House		0	0	0	

	Asian Restaurant	Athletics & Sports	Auditorium	Auto Dealership	\
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	

	BBQ Joint	Bakery	...	Toy / Game Store	Train Station	Tree	\
0	0	0	...	0	0	0	
1	0	0	...	0	0	0	
2	0	0	...	0	0	0	
3	0	0	...	0	0	0	
4	0	0	...	0	0	0	

	Vegetarian / Vegan Restaurant	Vietnamese Restaurant	Waterfront	Wine Bar	\
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	

	Wine Shop	Women's Store	Yoga Studio
0	0	0	0
1	0	0	0
2	0	0	0
3	0	0	0
4	0	0	0

[5 rows x 152 columns]

Next, let's group rows by institute and by taking the mean of the frequency of occurrence of each category

```
[35]: institutes_grouped = institutes_onehot.groupby('InstituteName').mean().
      ↪reset_index()
      institutes_grouped
```

[35]:

	InstituteName	Accessories Store \
0	ANNE'S Language House	0.00
1	American English Language School	0.00
2	Bouchereau Lingua International	0.00
3	Connect English Language Institute- Mission Va...	0.00
4	Connect English Language Institute- San Diego ...	0.00
5	Connect English- La Jolla	0.00
6	International Language Academy of Canada Vanco...	0.00
7	Language Studies International (LSI): Boston	0.00
8	Language Studies International (LSI): New York	0.01
9	Language Studies International (LSI): San Diego	0.00
10	Language Studies International (LSI): Toronto	0.00
11	Language Studies International (LSI): Vancouver	0.00
12	Metropolitan College	0.00

	American Restaurant	Art Gallery	Asian Restaurant	Athletics & Sports \
0	0.048780	0.024390	0.000000	0.000000
1	0.047619	0.000000	0.000000	0.000000
2	0.000000	0.000000	0.000000	0.000000
3	0.000000	0.000000	0.000000	0.000000
4	0.000000	0.000000	0.000000	0.000000
5	0.000000	0.000000	0.000000	0.000000
6	0.066667	0.000000	0.000000	0.000000
7	0.024390	0.000000	0.060976	0.000000
8	0.020000	0.000000	0.000000	0.000000
9	0.043478	0.000000	0.000000	0.000000
10	0.023810	0.000000	0.000000	0.02381
11	0.000000	0.011236	0.000000	0.000000
12	0.000000	0.000000	0.031250	0.000000

	Auditorium	Auto Dealership	BBQ Joint	Bakery ...	Toy / Game Store \
0	0.00	0.0000	0.00000	0.000000 ...	0.000000
1	0.00	0.0000	0.00000	0.000000 ...	0.000000
2	0.00	0.0000	0.00000	0.000000 ...	0.000000
3	0.00	0.0625	0.00000	0.000000 ...	0.000000
4	0.00	0.0000	0.00000	0.000000 ...	0.000000
5	0.00	0.0000	0.00000	0.000000 ...	0.000000
6	0.00	0.0000	0.00000	0.022222 ...	0.000000
7	0.00	0.0000	0.00000	0.097561 ...	0.000000
8	0.01	0.0000	0.01000	0.000000 ...	0.000000
9	0.00	0.0000	0.00000	0.000000 ...	0.000000
10	0.00	0.0000	0.02381	0.023810 ...	0.000000
11	0.00	0.0000	0.00000	0.044944 ...	0.022472
12	0.00	0.0000	0.00000	0.000000 ...	0.000000

	Train Station	Tree	Vegetarian / Vegan Restaurant	Vietnamese Restaurant \
0	0.0000	0.00	0.00000	0.024390

1	0.0000	0.00	0.00000	0.000000
2	0.0000	0.00	0.00000	0.000000
3	0.0625	0.00	0.00000	0.000000
4	0.0000	0.00	0.00000	0.000000
5	0.0000	0.00	0.00000	0.000000
6	0.0000	0.00	0.00000	0.000000
7	0.0000	0.00	0.02439	0.012195
8	0.0000	0.01	0.00000	0.000000
9	0.0000	0.00	0.00000	0.000000
10	0.0000	0.00	0.00000	0.000000
11	0.0000	0.00	0.00000	0.000000
12	0.0000	0.00	0.00000	0.000000

	Waterfront	Wine Bar	Wine Shop	Women's Store	Yoga Studio
0	0.000000	0.00000	0.000000	0.000000	0.000000
1	0.000000	0.00000	0.000000	0.047619	0.047619
2	0.000000	0.00000	0.000000	0.000000	0.000000
3	0.000000	0.00000	0.000000	0.000000	0.000000
4	0.000000	0.00000	0.000000	0.000000	0.000000
5	0.000000	0.00000	0.000000	0.000000	0.000000
6	0.022222	0.00000	0.000000	0.000000	0.000000
7	0.000000	0.00000	0.000000	0.000000	0.012195
8	0.000000	0.00000	0.030000	0.020000	0.000000
9	0.000000	0.00000	0.000000	0.000000	0.000000
10	0.000000	0.02381	0.000000	0.000000	0.023810
11	0.000000	0.00000	0.011236	0.000000	0.011236
12	0.000000	0.00000	0.000000	0.000000	0.000000

[13 rows x 152 columns]

Let's print each institute along with the top 5 most common venues

```
[36]: num_top_venues = 5

for institute in institutes_grouped['InstituteName']:
    print("----"+institute+"----")
    temp = institutes_grouped[institutes_grouped['InstituteName'] == institute].
    ↪T.reset_index()
    temp.columns = ['venue', 'freq']
    temp = temp.iloc[1:]
    temp['freq'] = temp['freq'].astype(float)
    temp = temp.round({'freq': 2})
    print(temp.sort_values('freq', ascending=False).reset_index(drop=True).
    ↪head(num_top_venues))
    print('\n')
```

```
----ANNE'S Language House----
      venue  freq
```


0	Restaurant	0.12
1	Hotel	0.10
2	Performing Arts Venue	0.07
3	Steakhouse	0.05
4	Coffee Shop	0.05

----American English Language School----

	venue	freq
0	Coffee Shop	0.10
1	Yoga Studio	0.05
2	Burger Joint	0.05
3	Food Truck	0.05
4	Gas Station	0.05

----Bouchereau Lingua International----

	venue	freq
0	Home Service	0.2
1	Diner	0.2
2	Chocolate Shop	0.2
3	Supermarket	0.2
4	Construction & Landscaping	0.2

----Connect English Language Institute- Mission Valley----

	venue	freq
0	Pub	0.06
1	Martial Arts School	0.06
2	Flower Shop	0.06
3	Performing Arts Venue	0.06
4	Rental Car Location	0.06

----Connect English Language Institute- San Diego (Pacific Beach Campus)----

	venue	freq
0	Intersection	0.15
1	Convenience Store	0.15
2	Hotel	0.15
3	Thai Restaurant	0.08
4	Nightclub	0.08

----Connect English- La Jolla----

	venue	freq
0	Park	0.33
1	Scenic Lookout	0.33
2	Golf Course	0.17

3	Sporting Goods Shop	0.17
4	Neighborhood	0.00

----International Language Academy of Canada Vancouver----

	venue	freq
0	Hotel	0.11
1	American Restaurant	0.07
2	Restaurant	0.07
3	Cosmetics Shop	0.04
4	Miscellaneous Shop	0.04

----Language Studies International (LSI): Boston----

	venue	freq
0	Bakery	0.10
1	Chinese Restaurant	0.10
2	Asian Restaurant	0.06
3	Coffee Shop	0.05
4	Food Truck	0.05

----Language Studies International (LSI): New York----

	venue	freq
0	Park	0.08
1	Pizza Place	0.06
2	Coffee Shop	0.04
3	Memorial Site	0.04
4	Hotel	0.04

----Language Studies International (LSI): San Diego----

	venue	freq
0	Park	0.09
1	Spa	0.09
2	Hotel	0.09
3	Middle Eastern Restaurant	0.04
4	Boutique	0.04

----Language Studies International (LSI): Toronto----

	venue	freq
0	Italian Restaurant	0.07
1	Coffee Shop	0.07
2	Café	0.05
3	Spa	0.05
4	Park	0.05

----Language Studies International (LSI): Vancouver----

	venue	freq
0	Hotel	0.06
1	Mexican Restaurant	0.04
2	Bakery	0.04
3	Japanese Restaurant	0.04
4	Concert Hall	0.03

----Metropolitan College----

	venue	freq
0	Pub	0.22
1	Coffee Shop	0.09
2	Italian Restaurant	0.06
3	Park	0.06
4	Sandwich Place	0.06

Let's put that into a *pandas* dataframe

First, let's write a function to sort the venues in descending order.

```
[37]: def return_most_common_venues(row, num_top_venues):
      row_categories = row.iloc[1:]
      row_categories_sorted = row_categories.sort_values(ascending=False)

      return row_categories_sorted.index.values[0:num_top_venues]
```

Now let's create the new dataframe and display the top 10 venues for each neighborhood.

```
[38]: num_top_venues = 5

      indicators = ['st', 'nd', 'rd']

      # create columns according to number of top venues
      columns = ['InstituteName']
      for ind in np.arange(num_top_venues):
          try:
              columns.append('{}{} Most Common Venue'.format(ind+1, indicators[ind]))
          except:
              columns.append('{}th Most Common Venue'.format(ind+1))

      # create a new dataframe
      institutes_venues_sorted = pd.DataFrame(columns=columns)
      institutes_venues_sorted['InstituteName'] = institutes_grouped['InstituteName']

      for ind in np.arange(institutes_grouped.shape[0]):
```

```

institutes_venues_sorted.iloc[ind, 1:] =
↪return_most_common_venues(institutes_grouped.iloc[ind, :], num_top_venues)

institutes_venues_sorted

```

```

[38]:
InstituteName 1st Most Common Venue \
0 ANNE'S Language House Restaurant
1 American English Language School Coffee Shop
2 Bouchereau Lingua International Diner
3 Connect English Language Institute- Mission Va... Martial Arts School
4 Connect English Language Institute- San Diego ... Convenience Store
5 Connect English- La Jolla Park
6 International Language Academy of Canada Vanco... Hotel
7 Language Studies International (LSI): Boston Bakery
8 Language Studies International (LSI): New York Park
9 Language Studies International (LSI): San Diego Hotel
10 Language Studies International (LSI): Toronto Italian Restaurant
11 Language Studies International (LSI): Vancouver Hotel
12 Metropolitan College Pub

2nd Most Common Venue 3rd Most Common Venue 4th Most Common Venue \
0 Hotel Performing Arts Venue Coffee Shop
1 Yoga Studio Gym / Fitness Center Food Truck
2 Home Service Supermarket Chocolate Shop
3 Gym Climbing Gym Clothing Store
4 Hotel Intersection Thai Restaurant
5 Scenic Lookout Sporting Goods Shop Golf Course
6 Restaurant American Restaurant Cosmetics Shop
7 Chinese Restaurant Asian Restaurant Coffee Shop
8 Pizza Place Coffee Shop Memorial Site
9 Park Spa Boutique
10 Coffee Shop Park Spa
11 Bakery Japanese Restaurant Mexican Restaurant
12 Coffee Shop Sandwich Place Italian Restaurant

5th Most Common Venue
0 Theater
1 Gas Station
2 Construction & Landscaping
3 Rental Car Location
4 Marijuana Dispensary
5 Flower Shop
6 Steakhouse
7 Food Truck
8 Hotel
9 Deli / Bodega
10 Café

```

11 French Restaurant
12 Park

7 Methodology: K-means

Run k -means to cluster the institutes into clusters.

```
[39]: # Create range of K values

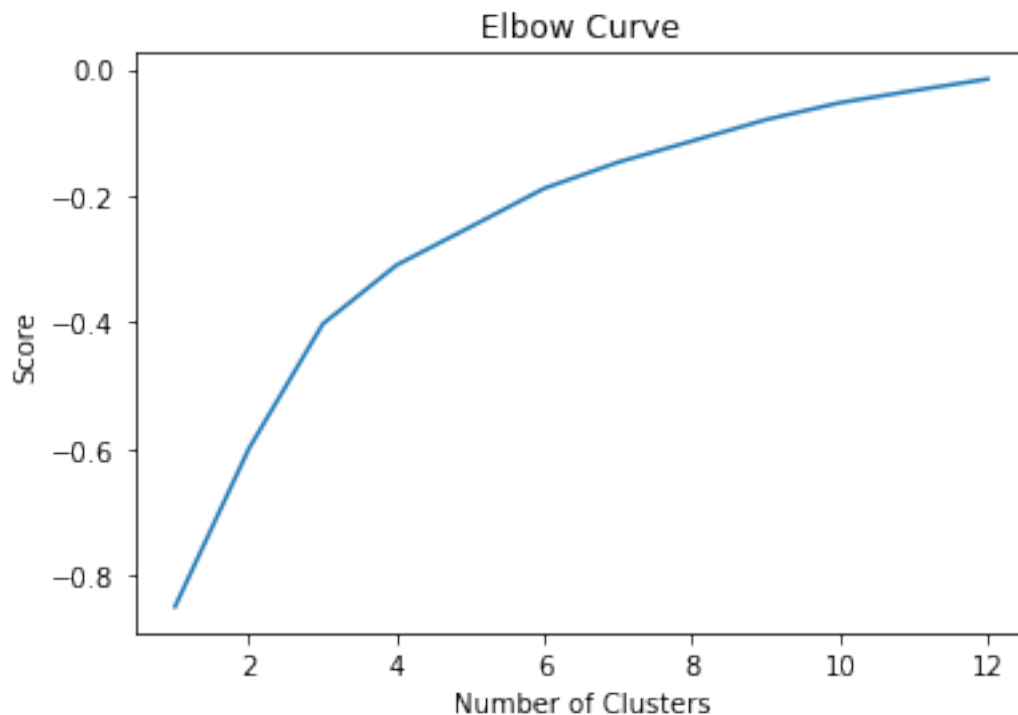
X = institutes_grouped_clustering = institutes_grouped.drop('InstituteName', 1)

Nc = range(1, 13)

#Create instances
kmeans = [KMeans(n_clusters=i, n_init = 12) for i in Nc]

#Fit the model and get scores for different K values
score = [kmeans[i].fit(X).score(X) for i in range(len(kmeans))]

#Plot the Elbow curve
plt.plot(Nc,score)
plt.xlabel('Number of Clusters')
plt.ylabel('Score')
plt.title('Elbow Curve')
plt.show()
```



```
[40]: # Let's select k as 3 because is where an elbow can be identified in the above
      ↪graph
      kclusters = 3

      institutes_grouped_clustering = institutes_grouped.drop('InstituteName', 1)

      # run k-means clustering
      kmeans = KMeans(n_clusters=kclusters, random_state=0).
      ↪fit(institutes_grouped_clustering)

      # check cluster labels generated for each row in the dataframe
      kmeans.labels_[0:10]
```

```
[40]: array([0, 0, 1, 0, 0, 2, 0, 0, 0, 0])
```

Let's create a new dataframe that includes the cluster as well as the top 10 venues for each institute.

```
[44]: # add clustering labels
      institutes_venues_sorted.insert(0, 'Cluster Labels', kmeans.labels_)

      institutes_merged = df_institutes

      # merge institutes_grouped with manhattan_data to add latitude/longitude for
      ↪each neighborhood
      institutes_merged = institutes_merged.join(institutes_venues_sorted.
      ↪set_index('InstituteName'), on='InstituteName')

      institutes_merged.drop(['index'], axis=1, inplace=True) # check the last
      ↪columns!
      institutes_merged
```

```
[44]:
```

	InstituteName	latitude	longitude	\
0	ANNE'S Language House	51.047150	-114.063340	
1	Metropolitan College	43.252430	-79.871300	
2	Language Studies International (LSI): Toronto	43.678510	-79.389730	
3	Language Studies International (LSI): Vancouver	49.280040	-123.124910	
4	International Language Academy of Canada Vanco...	49.288360	-123.122590	
5	Bouchereau Lingua International	45.466221	-74.077549	
6	Connect English Language Institute- San Diego ...	32.803720	-117.214110	
7	Connect English- La Jolla	32.852520	-117.186220	
8	Connect English Language Institute- Mission Va...	32.780980	-117.096280	
9	American English Language School	33.909860	-117.854170	
10	Language Studies International (LSI): Boston	42.350720	-71.058260	
11	Language Studies International (LSI): New York	40.708630	-74.014660	
12	Language Studies International (LSI): San Diego	32.723270	-117.160510	

	Country	Cluster Labels	1st Most Common Venue	2nd Most Common Venue	\
0	Canada	0	Restaurant	Hotel	
1	Canada	0	Pub	Coffee Shop	
2	Canada	0	Italian Restaurant	Coffee Shop	
3	Canada	0	Hotel	Bakery	
4	Canada	0	Hotel	Restaurant	
5	Canada	1	Diner	Home Service	
6	USA	0	Convenience Store	Hotel	
7	USA	2	Park	Scenic Lookout	
8	USA	0	Martial Arts School	Gym	
9	USA	0	Coffee Shop	Yoga Studio	
10	USA	0	Bakery	Chinese Restaurant	
11	USA	0	Park	Pizza Place	
12	USA	0	Hotel	Park	

	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
0	Performing Arts Venue	Coffee Shop	Theater
1	Sandwich Place	Italian Restaurant	Park
2	Park	Spa	Café
3	Japanese Restaurant	Mexican Restaurant	French Restaurant
4	American Restaurant	Cosmetics Shop	Steakhouse
5	Supermarket	Chocolate Shop	Construction & Landscaping
6	Intersection	Thai Restaurant	Marijuana Dispensary
7	Sporting Goods Shop	Golf Course	Flower Shop
8	Climbing Gym	Clothing Store	Rental Car Location
9	Gym / Fitness Center	Food Truck	Gas Station
10	Asian Restaurant	Coffee Shop	Food Truck
11	Coffee Shop	Memorial Site	Hotel
12	Spa	Boutique	Deli / Bodega

Finally, let's visualize the resulting clusters

```
[45]: institutes_merged["latitude"] = institutes_merged["latitude"].astype("float")
institutes_merged["longitude"] = institutes_merged["longitude"].astype("float")
```

```
[46]: # create map
map_clusters = folium.Map(location=[latitude, longitude], zoom_start=4)

# set color scheme for the clusters
x = np.arange(kclusters)
ys = [i + x + (i*x)**2 for i in range(kclusters)]
colors_array = cm.rainbow(np.linspace(0, 1, len(ys)))
rainbow = [colors.rgb2hex(i) for i in colors_array]

# add markers to the map
markers_colors = []
```

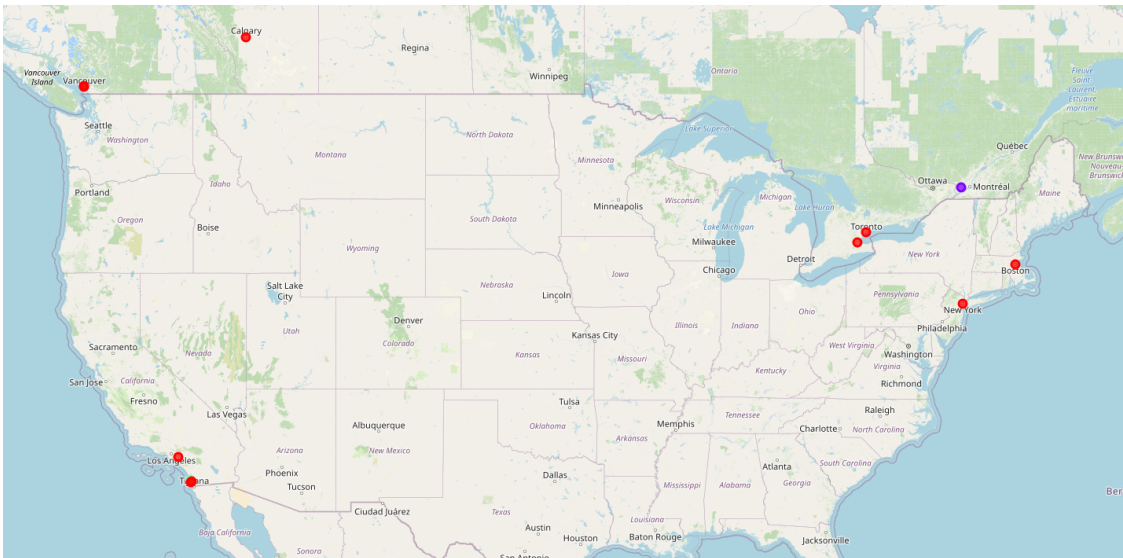
```

for lat, lon, poi, cluster in zip(institutes_merged['latitude'],
↳institutes_merged['longitude'], institutes_merged['InstituteName'],
↳institutes_merged['Cluster Labels']):
    label = folium.Popup(str(poi) + ' Cluster ' + str(cluster), parse_html=True)
    folium.CircleMarker(
        [lat, lon],
        radius=5,
        popup=label,
        color=rainbow[cluster-1],
        fill=True,
        fill_color=rainbow[cluster-1],
        fill_opacity=0.7).add_to(map_clusters)

map_clusters
#Image(filename='mapClusters.png')

```

[46]:



8 Results

Now, is possible to examine each cluster and determine the discriminating venue categories that distinguish each cluster.

```

[47]: institutes_merged.loc[institutes_merged['Cluster Labels'] == 0,
↳institutes_merged.columns[[1] + list(range(5, institutes_merged.shape[1]))]]

```

```

[47]:    latitude 1st Most Common Venue 2nd Most Common Venue \
0    51.04715          Restaurant          Hotel
1    43.25243              Pub          Coffee Shop
2    43.67851  Italian Restaurant          Coffee Shop

```


3	49.28004	Hotel	Bakery
4	49.28836	Hotel	Restaurant
6	32.80372	Convenience Store	Hotel
8	32.78098	Martial Arts School	Gym
9	33.90986	Coffee Shop	Yoga Studio
10	42.35072	Bakery	Chinese Restaurant
11	40.70863	Park	Pizza Place
12	32.72327	Hotel	Park

	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
0	Performing Arts Venue	Coffee Shop	Theater
1	Sandwich Place	Italian Restaurant	Park
2	Park	Spa	Café
3	Japanese Restaurant	Mexican Restaurant	French Restaurant
4	American Restaurant	Cosmetics Shop	Steakhouse
6	Intersection	Thai Restaurant	Marijuana Dispensary
8	Climbing Gym	Clothing Store	Rental Car Location
9	Gym / Fitness Center	Food Truck	Gas Station
10	Asian Restaurant	Coffee Shop	Food Truck
11	Coffee Shop	Memorial Site	Hotel
12	Spa	Boutique	Deli / Bodega

```
[48]: institutes_merged.loc[institutes_merged['Cluster Labels'] == 1,
    ↪institutes_merged.columns[[1] + list(range(5, institutes_merged.shape[1]))]]
```

```
[48]: latitude 1st Most Common Venue 2nd Most Common Venue \
5 45.466221 Diner Home Service

3rd Most Common Venue 4th Most Common Venue 5th Most Common Venue
5 Supermarket Chocolate Shop Construction & Landscaping
```

```
[49]: institutes_merged.loc[institutes_merged['Cluster Labels'] == 2,
    ↪institutes_merged.columns[[1] + list(range(5, institutes_merged.shape[1]))]]
```

```
[49]: latitude 1st Most Common Venue 2nd Most Common Venue 3rd Most Common Venue \
7 32.85252 Park Scenic Lookout Sporting Goods Shop

4th Most Common Venue 5th Most Common Venue
7 Golf Course Flower Shop
```

The 3 clusters according to the most common categories of venues for each group could be classified as:

Name	Cluster
LEISURE, STUDENTS, TOURISTS	0
FAMILIAR	1
NATURE LOVERS	2

Now that each institute has its own cluster, would be great to have the information about the courses that the institutes offers with their respective prices

This would help in some way to decide which course could be the best fit taking into account the venues, type of course and prices. The decision is up to the “new student”

```
[50]: institutes_all = df_courses.drop(columns=['latitude', 'longitude', 'Country',
↳ 'urlcourse', 'urlinstitute']) # Let's drop some columns because they are
↳ present in the table that is going to be merged with

institutes_all = institutes_merged.join(institutes_all.
↳ set_index('InstituteName'), on='InstituteName').drop(columns=['latitude',
↳ 'longitude']).reset_index().drop(columns=['index']) # Here is where the
↳ merge is done

institutes_all['PriceTo']=np.
↳ where(institutes_all['PriceTo']=='None',institutes_all['PriceFrom'],
↳ institutes_all['PriceTo']) # Let's replace the none values of the PriceTo
↳ column with the PriceFrom column
```

```
[51]: institutes_all["PriceTo"] = institutes_all["PriceTo"].astype("float")
institutes_all["PriceFrom"] = institutes_all["PriceFrom"].astype("float")

institutes_all
```

```
[51]:
```

	InstituteName	Country	Cluster	Labels \
0	ANNE'S Language House	Canada	0	
1	ANNE'S Language House	Canada	0	
2	Metropolitan College	Canada	0	
3	Metropolitan College	Canada	0	
4	Language Studies International (LSI): Toronto	Canada	0	
5	Language Studies International (LSI): Toronto	Canada	0	
6	Language Studies International (LSI): Toronto	Canada	0	
7	Language Studies International (LSI): Toronto	Canada	0	
8	Language Studies International (LSI): Vancouver	Canada	0	
9	Language Studies International (LSI): Vancouver	Canada	0	
10	Language Studies International (LSI): Vancouver	Canada	0	
11	Language Studies International (LSI): Vancouver	Canada	0	
12	Language Studies International (LSI): Vancouver	Canada	0	
13	International Language Academy of Canada Vanco...	Canada	0	
14	Bouchereau Lingua International	Canada	1	
15	Connect English Language Institute- San Diego ...	USA	0	
16	Connect English- La Jolla	USA	2	
17	Connect English- La Jolla	USA	2	
18	Connect English Language Institute- Mission Va...	USA	0	
19	Connect English Language Institute- Mission Va...	USA	0	
20	American English Language School	USA	0	
21	Language Studies International (LSI): Boston	USA	0	

22	Language Studies International (LSI): Boston	USA	0
23	Language Studies International (LSI): Boston	USA	0
24	Language Studies International (LSI): Boston	USA	0
25	Language Studies International (LSI): New York	USA	0
26	Language Studies International (LSI): New York	USA	0
27	Language Studies International (LSI): San Diego	USA	0
28	Language Studies International (LSI): San Diego	USA	0
29	Language Studies International (LSI): San Diego	USA	0

	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue \
0	Restaurant	Hotel	Performing Arts Venue
1	Restaurant	Hotel	Performing Arts Venue
2	Pub	Coffee Shop	Sandwich Place
3	Pub	Coffee Shop	Sandwich Place
4	Italian Restaurant	Coffee Shop	Park
5	Italian Restaurant	Coffee Shop	Park
6	Italian Restaurant	Coffee Shop	Park
7	Italian Restaurant	Coffee Shop	Park
8	Hotel	Bakery	Japanese Restaurant
9	Hotel	Bakery	Japanese Restaurant
10	Hotel	Bakery	Japanese Restaurant
11	Hotel	Bakery	Japanese Restaurant
12	Hotel	Bakery	Japanese Restaurant
13	Hotel	Restaurant	American Restaurant
14	Diner	Home Service	Supermarket
15	Convenience Store	Hotel	Intersection
16	Park	Scenic Lookout	Sporting Goods Shop
17	Park	Scenic Lookout	Sporting Goods Shop
18	Martial Arts School	Gym	Climbing Gym
19	Martial Arts School	Gym	Climbing Gym
20	Coffee Shop	Yoga Studio	Gym / Fitness Center
21	Bakery	Chinese Restaurant	Asian Restaurant
22	Bakery	Chinese Restaurant	Asian Restaurant
23	Bakery	Chinese Restaurant	Asian Restaurant
24	Bakery	Chinese Restaurant	Asian Restaurant
25	Park	Pizza Place	Coffee Shop
26	Park	Pizza Place	Coffee Shop
27	Hotel	Park	Spa
28	Hotel	Park	Spa
29	Hotel	Park	Spa

	4th Most Common Venue	5th Most Common Venue	Location \
0	Coffee Shop	Theater	Calgary, Canada
1	Coffee Shop	Theater	Calgary, Canada
2	Italian Restaurant	Park	Hamilton, Canada
3	Italian Restaurant	Park	Hamilton, Canada
4	Spa	Café	Toronto, Canada

5	Spa	Café	Toronto, Canada
6	Spa	Café	Toronto, Canada
7	Spa	Café	Toronto, Canada
8	Mexican Restaurant	French Restaurant	Vancouver, Canada
9	Mexican Restaurant	French Restaurant	Vancouver, Canada
10	Mexican Restaurant	French Restaurant	Vancouver, Canada
11	Mexican Restaurant	French Restaurant	Vancouver, Canada
12	Mexican Restaurant	French Restaurant	Vancouver, Canada
13	Cosmetics Shop	Steakhouse	Vancouver, Canada
14	Chocolate Shop	Construction & Landscaping	Montreal, Canada
15	Thai Restaurant	Marijuana Dispensary	San Diego, USA
16	Golf Course	Flower Shop	San Diego, USA
17	Golf Course	Flower Shop	San Diego, USA
18	Clothing Store	Rental Car Location	San Diego, USA
19	Clothing Store	Rental Car Location	San Diego, USA
20	Food Truck	Gas Station	Los Angeles, USA
21	Coffee Shop	Food Truck	Boston, USA
22	Coffee Shop	Food Truck	Boston, USA
23	Coffee Shop	Food Truck	Boston, USA
24	Coffee Shop	Food Truck	Boston, USA
25	Memorial Site	Hotel	New York City, USA
26	Memorial Site	Hotel	New York City, USA
27	Boutique	Deli / Bodega	San Diego, USA
28	Boutique	Deli / Bodega	San Diego, USA
29	Boutique	Deli / Bodega	San Diego, USA

	CourseName	PriceFrom	PriceTo	\
0	General English 20 (GE20)	354.0	687.0	
1	General English (GE25)	384.0	717.0	
2	ESL Private Lessons	384.0	619.0	
3	Intensive English	290.0	525.0	
4	Intensive 30	495.0	823.0	
5	General 20	431.0	759.0	
6	Club 40+ (20 lessons per week plus afternoon a...	708.0	1037.0	
7	Afternoon 10	286.0	614.0	
8	General 20	431.0	759.0	
9	Intensive 30	495.0	823.0	
10	Intensive 25	465.0	794.0	
11	Club 40+ (20 lessons per week plus afternoon a...	708.0	1037.0	
12	Afternoon 10	286.0	614.0	
13	General English - Intensive	503.0	870.0	
14	Super Intensive English	533.0	887.0	
15	English Max	390.0	390.0	
16	English Focus Course (12 hrs/week)	365.0	365.0	
17	English Max Course (18 hrs/week)	390.0	390.0	
18	English Max Course (18 hrs/week)	390.0	390.0	
19	English Focus Course (12 hrs/week)	315.0	315.0	

20	ESL Program (12 Weeks Minimum)	335.0	335.0
21	Intensive 30	645.0	960.0
22	Club 40+ (20 lessons per week plus afternoon a...	840.0	1155.0
23	One-to-One (5 lessons per week)	575.0	890.0
24	General 20	535.0	850.0
25	One-to-One (20 lessons per week)	1955.0	2280.0
26	One-to-One (10 lessons per week)	1055.0	1380.0
27	Intensive 30	645.0	950.0
28	Club 40+ (20 lessons per week plus afternoon a...	840.0	1240.0
29	General 20	535.0	935.0

	Address
0	101 6th Avenue S.W., Suite 1250, Calgary, Albe...
1	101 6th Avenue S.W., Suite 1250, Calgary, Albe...
2	146 James Street South, Hamilton, Ontario,, Ha...
3	146 James Street South, Hamilton, Ontario,, Ha...
4	1055 Yonge Street, Suite #210, Toronto, ON M4W...
5	1055 Yonge Street, Suite #210, Toronto, ON M4W...
6	1055 Yonge Street, Suite #210, Toronto, ON M4W...
7	1055 Yonge Street, Suite #210, Toronto, ON M4W...
8	101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...
9	101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...
10	101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...
11	101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...
12	101-808 Nelson Street, Vancouver, BC V6Z 2H2, ...
13	1199 West Pender Street, Vancouver, British Co...
14	70 Notre Dame West, Suite 400, Montréal, Quebe...
15	3565 Del Rey St., Suite 300, San Diego, Califo...
16	5090 Shoreham Place, Suite 206, San Diego, Cal...
17	5090 Shoreham Place, Suite 206, San Diego, Cal...
18	4560 Alvarado Canyon Road, Suite 2B, San Diego...
19	4560 Alvarado Canyon Road, Suite 2B, San Diego...
20	3230 E. Imperial Hwy, Suite 301, Brea, Califor...
21	105 Beach Street, Boston, MA 02111, USA
22	105 Beach Street, Boston, MA 02111, USA
23	105 Beach Street, Boston, MA 02111, USA
24	105 Beach Street, Boston, MA 02111, USA
25	40 Rector Street, 10th Floor, Suite 1000, New ...
26	40 Rector Street, 10th Floor, Suite 1000, New ...
27	1706 5th Avenue, San Diego, CA 92101, USA
28	1706 5th Avenue, San Diego, CA 92101, USA
29	1706 5th Avenue, San Diego, CA 92101, USA

Let's calculate the average price from and to and the general as well

[52]:

```

institutes_prices = institutes_all.groupby(['InstituteName'])['Cluster Labels',
↳ 'PriceFrom', 'PriceTo'].mean().reset_index() #Let's calculate the mean of
↳ the prices by column
institutes_prices['PriceAvg'] = (institutes_prices['PriceFrom'] +
↳ institutes_prices['PriceTo']) / 2 #Let's calculate the mean of the prices by
↳ row

institutes_prices = institutes_prices.sort_values('PriceAvg', ascending=False).
↳ reset_index(drop=True)

institutes_prices['Cluster Labels'] = np.where(institutes_prices['Cluster
↳ Labels'] == 0, 'LEISURE, STUDENTS, TOURISTS', np.
↳ where(institutes_prices['Cluster Labels'] == 1, 'FAMILIAR', 'NATURE LOVERS'))

institutes_prices

```

<ipython-input-52-a505625daf17>:1: FutureWarning: Indexing with multiple keys (implicitly converted to a tuple of keys) will be deprecated, use a list instead.

```

institutes_prices = institutes_all.groupby(['InstituteName'])['Cluster
Labels', 'PriceFrom', 'PriceTo'].mean().reset_index() #Let's calculate the mean
of the prices by column

```

```

[52]:
InstituteName \
0      Language Studies International (LSI): New York
1      Language Studies International (LSI): San Diego
2      Language Studies International (LSI): Boston
3      Bouchereau Lingua International
4      International Language Academy of Canada Vanco...
5      Language Studies International (LSI): Toronto
6      Language Studies International (LSI): Vancouver
7      ANNE'S Language House
8      Metropolitan College
9      Connect English Language Institute- San Diego ...
10     Connect English- La Jolla
11     Connect English Language Institute- Mission Va...
12     American English Language School

Cluster Labels  PriceFrom  PriceTo  PriceAvg
0  LEISURE, STUDENTS, TOURISTS  1505.000000  1830.000000  1667.500
1  LEISURE, STUDENTS, TOURISTS  673.333333  1041.666667  857.500
2  LEISURE, STUDENTS, TOURISTS  648.750000  963.750000  806.250
3      FAMILIAR  533.000000  887.000000  710.000
4  LEISURE, STUDENTS, TOURISTS  503.000000  870.000000  686.500
5  LEISURE, STUDENTS, TOURISTS  480.000000  808.250000  644.125
6  LEISURE, STUDENTS, TOURISTS  477.000000  805.400000  641.200
7  LEISURE, STUDENTS, TOURISTS  369.000000  702.000000  535.500

```

8	LEISURE, STUDENTS, TOURISTS	337.000000	572.000000	454.500
9	LEISURE, STUDENTS, TOURISTS	390.000000	390.000000	390.000
10	NATURE LOVERS	377.500000	377.500000	377.500
11	LEISURE, STUDENTS, TOURISTS	352.500000	352.500000	352.500
12	LEISURE, STUDENTS, TOURISTS	335.000000	335.000000	335.000

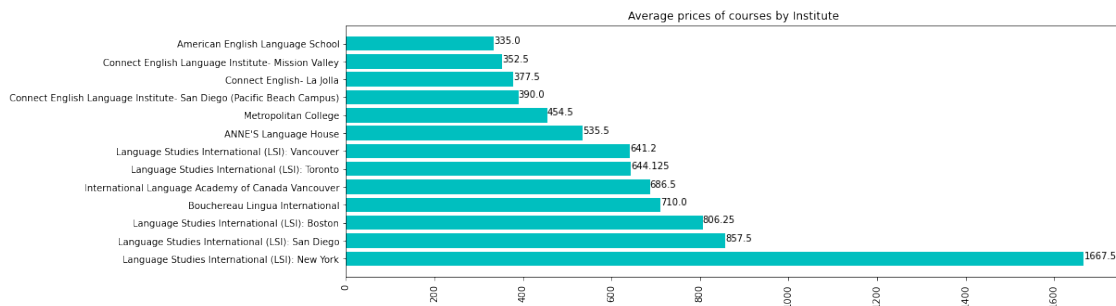
```
[53]: fig, ax = plt.subplots(figsize=(15, 5))

plt.title('Average prices of courses by Institute')
plt.xticks(rotation=90)

ax.barh(institutes_prices['InstituteName'],institutes_prices['PriceAvg'],
        color='c')

for index, value in enumerate(institutes_prices['PriceAvg']):
    plt.text(value, index, str(value))

plt.show()
print('Done!')
```



Done!

9 Discussion

This analysis shows that from the 30 courses selected from Canada and USA, 13 institutes are offering them in different locations of the countries. The locations are actually quite far one from each other (except for San Diego, USA where 4 courses are relatively closer). In this sense, takes a lot of importance the decision of selecting which course to take according to the personal “taste”

The venues around the institutes, after getting the data from the Foursquare API, **151** unique categories were identified. It is good to know that the radius that was selected for the API was 500 meters that is a distance considerably closer from the institutes.

The point here is that there are only 3 clusters created and one of them has the majority of institutes. After looking at the categories of the venues from each cluster, the possible classification can be:

- LEISURE, STUDENTS, TOURISTS

- FAMILIAR
- NATURE LOVERS

The variety of categories of venues around the main one **LEISURE, STUDENTS, TOURISTS** let us assume that the public could be really diverse whereas in the other clusters, a more specific “taste” is found. It means, that for people for example that is thinking to travel with their families or that likes a lot the nature could be quite easier to take a decision. However, it is good to highlight that according to the locations of the institutes, there is a big opportunity of improvement of this model in terms of adding different variables that includes for example weather, temperature or cost of living. I would say that this is just a general idea of what to expect around the locations of study.

Finally, the costs should be taken into account because the average of the courses are between 335 USD and 1667 USD which let us think that other variables such as reviews, ratings, comments or duration would be great to include to have an explanation of why the prices are so diverse.

10 Conclusion

This can be the first approach and as it is stated above, there is some additional information that can be useful to allow the “new student” to take a better decision. The good point here, is that, we already have a general view of the courses offered by the best institutes according to the website <https://www.languageinternational.com/> clustered by the categories of venues found around them.

[]: