

# PalmBeachCounty

April 21, 2020

New Construction and Potential for New Businesses in Neighborhoods in Palm Beach County

## 0.1 Introduction

In this report, we take data from Palm Beach County Planning, Zoning and Building, 2019 Building Permit Reports, (found here <http://discover.pbcgov.org/pzb/planning/Pages/Permit-Activity-Reports.aspx> ) to find where the county is permitting new housing construction. Also, we will use the Foursquare API to explore neighborhoods in Palm Beach County. We will do analysis to determine what venues maybe needed to service these new planned housing developments. We will use the **explore** function to get the most common venue categories in each neighborhood, and then use this feature to group the neighborhoods into clusters. You will use the  $k$ -means clustering algorithm to complete this task. Finally, we will use the Folium library to visualize the neighborhoods in Palm Beach County and their emerging clusters.

## 0.2 Table of Contents

1. Download and Explore Dataset
2. Explore Neighborhoods in Palm Beach County
3. Analyze Each Neighborhood
4. Cluster Neighborhoods
5. Examine Clusters

Import libraries

```
[162]: {
    "tags": [
        "remove_output",
    ]
}

import requests # library to handle requests
import pandas as pd # library for data analysis
import numpy as np # library to handle data in a vectorized manner
import json # library to handle json files
import random # library for random number generation
```

```

#scraping pdf
!pip install tabula-py
!pip install tabulate
import tabula
import tabulate

#excell files
import xlrd

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

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

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

#libraries for displaying images
from IPython.display import Image
from IPython.core.display import HTML

#transforming json file into a pandas dataframe library
from pandas.io.json import json_normalize

!conda install -c conda-forge folium=0.5.0 --yes
import folium # plotting library

print('Folium installed')
print('Libraries imported.')

```

Requirement already satisfied: tabula-py in  
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (2.1.0)

Requirement already satisfied: distro in  
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from tabula-py)  
(1.5.0)

Requirement already satisfied: pandas>=0.25.3 in  
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from tabula-py)  
(1.0.3)

Requirement already satisfied: numpy in  
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from tabula-py)  
(1.18.1)

```
Requirement already satisfied: pytz>=2017.2 in
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from
pandas>=0.25.3->tabula-py) (2019.3)
Requirement already satisfied: python-dateutil>=2.6.1 in
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from
pandas>=0.25.3->tabula-py) (2.8.1)
Requirement already satisfied: six>=1.5 in
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from python-
dateutil>=2.6.1->pandas>=0.25.3->tabula-py) (1.14.0)
Requirement already satisfied: tabulate in
/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (0.8.7)
Collecting package metadata (current_repodata.json): done
Solving environment: done
```

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

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

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

```
Folium installed
Libraries imported.
```

```
[2]: import os
      print (os.getcwd())
```

```
/resources/labs/DP0701EN
```

We created a new excel dataframe which coordinates the Cities in the Municipality columns with their coordinates found by this website [www.lat-long.com](http://www.lat-long.com) or <https://www.findlatitudeandlongitude.com/> or [latlong.net](http://latlong.net)

```
[3]: lat_lon = pd.read_excel(r'/resources/labs/DP0701EN/Palm Beach County Cities and
      ↪Zips.xlsx')
      lat_lon.head()
```

```
[3]:
```

	MUNICIPALITIES	Latitude	Longitude
0	Atlantis	26.590902	-80.100876
1	Belle Glade	26.684510	-80.667558
2	Boca Raton	26.358688	-80.083098
3	Boynton Beach	26.525349	-80.066431
4	Briny Breezes town	26.508405	-80.050875

For the pdf dataframe ... we used the tabula technology to extract the dataframe to a csv file <https://tabula.technology/>

```
[4]: PAR = pd.read_csv(r'/resources/labs/DP0701EN/
↳tabula-4thQuarterPermitActivityReport.csv', sep=',', header=None, names =
↳["MUNICIPALITIES", "SINGLE FAMILY UNITS", "SFU VALUE", "MULTI FAMILY UNITS",
↳"MFU VALUE", "TOTAL UNITS", "TOTAL UNITS VALUE"])
PAR.head()
```

```
[4]:
```

	MUNICIPALITIES	SINGLE FAMILY UNITS	SFU VALUE	MULTI FAMILY UNITS	\
0	Atlantis	1	\$720,000	0	
1	Belle Glade	2	\$300,000	0	
2	Boca Raton	26	\$20,294,965	5	
3	Boynton Beach	3	\$712,370	116	
4	Briny Breezes town	0	\$0	0	

	MFU VALUE	TOTAL UNITS	TOTAL UNITS VALUE
0	\$0	1	\$720,000
1	\$0	2	\$300,000
2	\$576,252	31	\$20,871,217
3	\$6,217,584	119	\$6,929,954
4	\$0	0	\$0

```
[5]: #sorting values by total units and take the top 5 Municipalities of new
↳construction to look at.
Top5 =PAR.sort_values(by = 'TOTAL UNITS', ascending = False).head()
Top5
```

```
[5]:
```

	MUNICIPALITIES	SINGLE FAMILY UNITS	SFU VALUE	\
26	Palm Beach County Unincorporated Area	543	\$191,834,714	
37	West Palm Beach	17	\$4,036,764	
3	Boynton Beach	3	\$712,370	
38	Westlake	76	\$17,596,144	
27	Palm Beach Gardens	53	\$24,789,011	

	MULTI FAMILY UNITS	MFU VALUE	TOTAL UNITS	TOTAL UNITS VALUE
26	156	\$8,566,105	699	\$200,400,819
37	558	\$56,036,486	575	\$60,073,250
3	116	\$6,217,584	119	\$6,929,954
38	0	\$0	76	\$17,596,144
27	20	\$1,981,103	73	\$26,770,114

merge dataframes

```
[6]: #merge dataframes
PBC_df= pd.merge( Top5,lat_lon, on='MUNICIPALITIES')
PBC_df
```

```
[6]:
```

	MUNICIPALITIES	SINGLE FAMILY UNITS	SFU VALUE	\
0	Palm Beach County Unincorporated Area	543	\$191,834,714	
1	West Palm Beach	17	\$4,036,764	
2	Boynton Beach	3	\$712,370	
3	Westlake	76	\$17,596,144	
4	Palm Beach Gardens	53	\$24,789,011	

	MULTI FAMILY UNITS	MFU VALUE	TOTAL UNITS	TOTAL UNITS VALUE	Latitude	\
0	156	\$8,566,105	699	\$200,400,819	26.709723	
1	558	\$56,036,486	575	\$60,073,250	26.715300	
2	116	\$6,217,584	119	\$6,929,954	26.525349	
3	0	\$0	76	\$17,596,144	26.747000	
4	20	\$1,981,103	73	\$26,770,114	26.839600	

```

Longitude
0 -80.064163
1 -80.053400
2 -80.066431
3 -80.307700
4 -80.101900

```

## EXPLORE AND CLUSTER

```
[7]: ## MAP OF TOP 5 Municipalities of new construction
```

```
[8]: address = 'Palm Beach County'

geolocator = Nominatim(user_agent="PBC_explorer")
location = geolocator.geocode(address)
latitude = location.latitude
longitude = location.longitude
print('The geograpical coordinate of Palm Beach County are {}, {}'.format(
    latitude, longitude))
```

The geograpical coordinate of Palm Beach County are 26.6279798, -80.4494174.

```
[9]: map_PBC = folium.Map(location=[latitude,longitude],zoom_start=10)

for lat,lng,municipality in zip(PBC_df['Latitude'],PBC_df['Longitude'],PBC_df['MUNICIPALITIES']):
    label = '{}'.format(municipality)
    label = folium.Popup(label, parse_html=True)
    folium.CircleMarker(
        [lat,lng],
        radius=2,
        popup=label,
        color='blue',
        fill=True,
```

```

        fill_color='#3186cc',
        fill_opacity=0.7,
        parse_html=False).add_to(map_PBC)
map_PBC

```

[9]: <folium.folium.Map at 0x7f423883bb00>

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

### 0.3 Define Foursquare Credentials and Version

```

[10]: CLIENT_ID = 'LAUAGY5VQH2DJ4VUXN40XSN GEGCKTOTLXSDAS04FL1XB4SES' # your_
      ↪ Foursquare ID
CLIENT_SECRET = 'D3CTWFSGB2D5XWODQ02ZB2VGVOI2IZMIOISACKDSVCL0MEV2' # your_
      ↪ Foursquare Secret
VERSION = '20180605'

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

```

Your credentails:

CLIENT\_ID: LAUAGY5VQH2DJ4VUXN40XSN GEGCKTOTLXSDAS04FL1XB4SES

CLIENT\_SECRET: D3CTWFSGB2D5XWODQ02ZB2VGVOI2IZMIOISACKDSVCL0MEV2

### 0.4 Let's explore the neighborhood of Palm Beach County

get venues

```

[11]: LIMIT = 100 # limit of number of venues returned by Foursquare API
      radius = 100000 # define radius

```

```

[12]: # create URL
url = 'https://api.foursquare.com/v2/venues/explore?
      ↪ &client_id={} &client_secret={} &v={} &ll={},{} &radius={} &limit={}' .format(
      CLIENT_ID,
      CLIENT_SECRET,
      VERSION,
      latitude,
      longitude,
      radius,
      LIMIT)
url # display URL

```

```

[12]: 'https://api.foursquare.com/v2/venues/explore?&client_id=LAUAGY5VQH2DJ4VUXN40XSN
      GEGCKTOTLXSDAS04FL1XB4SES&client_secret=D3CTWFSGB2D5XWODQ02ZB2VGVOI2IZMIOISACKDS
      VCL0MEV2&v=20180605&ll=26.6279798,-80.4494174&radius=100000&limit=100'

```

send Get request to examn the results

```
[13]: results = requests.get(url).json()
```

```
[14]: # function that extracts the category of the venue
def get_category_type(row):
    try:
        categories_list = row['categories']
    except:
        categories_list = row['venue.categories']

    if len(categories_list) == 0:
        return None
    else:
        return categories_list[0]['name']
```

## 1 Clean Data in JSon File put it in panda dataframe

```
[15]: venues = results['response']['groups'][0]['items']

nearby_venues = json_normalize(venues) # flatten JSON

# filter columns
filtered_columns = ['venue.name', 'venue.categories', 'venue.location.lat',
                    ↪ 'venue.location.lng']
nearby_venues = nearby_venues.loc[:, filtered_columns]

# filter the category for each row
nearby_venues['venue.categories'] = nearby_venues.apply(get_category_type,
                    ↪ axis=1)

# clean columns
nearby_venues.columns = [col.split(".")[1] for col in nearby_venues.columns]

nearby_venues.head()
```

/home/jupyterlab/conda/envs/python/lib/python3.6/site-packages/ipykernel\_launcher.py:3: FutureWarning: pandas.io.json.json\_normalize is deprecated, use pandas.json\_normalize instead

This is separate from the ipykernel package so we can avoid doing imports until

```
[15]:
```

	name	categories	lat	\
0	Trader Joe's	Grocery Store	26.636809	
1	Whole Foods Market	Grocery Store	26.641351	
2	Chick-fil-A	Fast Food Restaurant	26.676292	

3	La Perrada del Gordo	Spanish Restaurant	26.638005
4	Morikami Museum And Japanese Gardens	Garden	26.428870

```

lng
0 -80.205553
1 -80.205912
2 -80.201740
3 -80.112046
4 -80.156608

```

```

[16]: 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 = ['MUNICIPALITIES',
                              'Municipality Latitude',
                              'Municipality Longitude',
                              'Venue',
                              'Venue Latitude',

```



```

        'Venue Longitude',
        'Venue Category']

    return(nearby_venues)

```

```

[17]: PBC_venues = getNearbyVenues(names=PBC_df['MUNICIPALITIES'],
                                   latitudes=PBC_df['Latitude'],
                                   longitudes=PBC_df['Longitude']
                                   )

```

Palm Beach County Unincorporated Area  
 West Palm Beach  
 Boynton Beach  
 Westlake  
 Palm Beach Gardens

Get new dataframe with venues

```

[18]: PBC_venues.head()

```

```

[18]:
      MUNICIPALITIES Municipality Latitude \
0  Palm Beach County Unincorporated Area    26.709723
1  Palm Beach County Unincorporated Area    26.709723
2  Palm Beach County Unincorporated Area    26.709723
3  Palm Beach County Unincorporated Area    26.709723
4  Palm Beach County Unincorporated Area    26.709723

      Municipality Longitude      Venue Venue Latitude \
0          -80.064163  Victoria's Secret PINK    26.706917
1          -80.064163  West Palm Beach Marriott    26.707069
2          -80.064163  Okeechobee & Parker RR Xing    26.706047
3          -80.064163    Bistro Ten Zero One    26.707419
4          -80.064163    Shoes For Crews LLC    26.712893

      Venue Longitude Venue Category
0          -80.060949  Lingerie Store
1          -80.063349      Hotel
2          -80.062406    Rest Area
3          -80.063645      Diner
4          -80.064453    Shoe Store

```

Group by Municipality

```

[19]: PBC_venues.groupby('MUNICIPALITIES').count().head()

```

```

[19]:
      MUNICIPALITIES Municipality Latitude \
Boynton Beach                                6

```

Palm Beach County Unincorporated Area	8
Palm Beach Gardens	20
West Palm Beach	71
Westlake	2

	Municipality Longitude	Venue \
MUNICIPALITIES		
Boynton Beach	6	6
Palm Beach County Unincorporated Area	8	8
Palm Beach Gardens	20	20
West Palm Beach	71	71
Westlake	2	2

	Venue Latitude	Venue Longitude \
MUNICIPALITIES		
Boynton Beach	6	6
Palm Beach County Unincorporated Area	8	8
Palm Beach Gardens	20	20
West Palm Beach	71	71
Westlake	2	2

	Venue Category
MUNICIPALITIES	
Boynton Beach	6
Palm Beach County Unincorporated Area	8
Palm Beach Gardens	20
West Palm Beach	71
Westlake	2

```
[20]: print('There are {} uniques categories.'.format(len(PBC_venues['Venue_
↳Category'].unique())))
```

There are 65 uniques categories.

Merge dataframes to get the new family units built by each Municipalities

```
[21]: NewHousingPBC = pd.merge(PAR, PBC_venues, on='MUNICIPALITIES')
NewHousingPBC.head()
```

	MUNICIPALITIES	SINGLE FAMILY UNITS	SFU VALUE	MULTI FAMILY UNITS \
0	Boynton Beach	3	\$712,370	116
1	Boynton Beach	3	\$712,370	116
2	Boynton Beach	3	\$712,370	116
3	Boynton Beach	3	\$712,370	116
4	Boynton Beach	3	\$712,370	116

	MFU VALUE	TOTAL UNITS	TOTAL UNITS VALUE	Municipality Latitude \
0	\$6,217,584	119	\$6,929,954	26.525349

1	\$6,217,584	119	\$6,929,954	26.525349
2	\$6,217,584	119	\$6,929,954	26.525349
3	\$6,217,584	119	\$6,929,954	26.525349
4	\$6,217,584	119	\$6,929,954	26.525349

	Municipality	Longitude	Venue \
0	-80.066431	Schoolhouse Children's Museum & Learning Center	
1	-80.066431	Sailfish Cafe	
2	-80.066431	South Beach Baby	
3	-80.066431	Concrete Solutions Fl	
4	-80.066431	Mulch Park	

	Venue Latitude	Venue Longitude	Venue Category
0	26.527577	-80.062953	Museum
1	26.526408	-80.063509	Café
2	26.523549	-80.063925	Beach
3	26.524458	-80.063360	Construction & Landscaping
4	26.526872	-80.062625	Playground

## 2 Analyze Each Neighborhood

```
[22]: # one hot encoding
PBC_onehot = pd.get_dummies(PBC_venues[['Venue Category']], prefix="",
                             ↪prefix_sep="")

# add neighborhood column back to dataframe
PBC_onehot['MUNICIPALITIES'] = PBC_venues['MUNICIPALITIES']

# move neighborhood column to the first column
fixed_columns = [PBC_onehot.columns[-1]] + list(PBC_onehot.columns[:-1])
PBC_onehot = PBC_onehot[fixed_columns]

PBC_onehot.head()
```

```
[22]: MUNICIPALITIES American Restaurant \
0 Palm Beach County Unincorporated Area 0
1 Palm Beach County Unincorporated Area 0
2 Palm Beach County Unincorporated Area 0
3 Palm Beach County Unincorporated Area 0
4 Palm Beach County Unincorporated Area 0

Asian Restaurant Bank Bar Baseball Field Beach Bistro Breakfast Spot \
0 0 0 0 0 0 0 0
1 0 0 0 0 0 0 0
2 0 0 0 0 0 0 0
3 0 0 0 0 0 0 0
```

4		0	0	0		0	0	0		0
---	--	---	---	---	--	---	---	---	--	---

	Brewery	...	Sandwich Place	Sculpture Garden	Shipping Store	Shoe Store	\
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	1	

	Shopping Mall	Steakhouse	Sushi Restaurant	Theater	Train Station	\
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	

	Wine Bar
0	0
1	0
2	0
3	0
4	0

[5 rows x 66 columns]

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

```
[23]: PBC_grouped = PBC_onehot.groupby('MUNICIPALITIES').mean().reset_index()
PBC_grouped.head()
```

```
[23]:
```

	MUNICIPALITIES	American Restaurant	\
0	Boynton Beach	0.166667	
1	Palm Beach County Unincorporated Area	0.000000	
2	Palm Beach Gardens	0.050000	
3	West Palm Beach	0.014085	
4	Westlake	0.000000	

	Asian Restaurant	Bank	Bar	Baseball Field	Beach	Bistro	\
0	0.000000	0.000000	0.000000	0.0	0.166667	0.000000	
1	0.000000	0.000000	0.000000	0.0	0.000000	0.000000	
2	0.000000	0.000000	0.000000	0.0	0.000000	0.000000	
3	0.042254	0.042254	0.056338	0.0	0.000000	0.014085	
4	0.000000	0.000000	0.000000	0.5	0.000000	0.000000	

	Breakfast Spot	Brewery	...	Sandwich Place	Sculpture Garden	\
0	0.000000	0.000000	...	0.000000	0.000000	

1	0.000000	0.000000	...	0.000000	0.000000
2	0.050000	0.000000	...	0.000000	0.000000
3	0.014085	0.014085	...	0.028169	0.014085
4	0.000000	0.000000	...	0.000000	0.000000

	Shipping Store	Shoe Store	Shopping Mall	Steakhouse	Sushi Restaurant \
0	0.00	0.000	0.00	0.00	0.000000
1	0.00	0.125	0.00	0.00	0.000000
2	0.05	0.050	0.05	0.05	0.050000
3	0.00	0.000	0.00	0.00	0.014085
4	0.00	0.000	0.00	0.00	0.000000

	Theater	Train Station	Wine Bar
0	0.000000	0.000000	0.000000
1	0.000000	0.000000	0.000000
2	0.000000	0.000000	0.000000
3	0.014085	0.014085	0.014085
4	0.000000	0.000000	0.000000

[5 rows x 66 columns]

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

```
[24]: num_top_venues = 5

for hood in PBC_grouped['MUNICIPALITIES']:
    print("----"+hood+"----")
    temp = PBC_grouped[PBC_grouped['MUNICIPALITIES'] == hood].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')
```

----Boynton Beach----

	venue	freq
0	American Restaurant	0.17
1	Beach	0.17
2	Museum	0.17
3	Playground	0.17
4	Café	0.17

----Palm Beach County Unincorporated Area----

	venue	freq
0	Bus Station	0.25

1	Diner	0.12
2	Hotel	0.12
3	Lingerie Store	0.12
4	Rest Area	0.12

----Palm Beach Gardens----

	venue	freq
0	Hotel	0.10
1	American Restaurant	0.05
2	Donut Shop	0.05
3	Sushi Restaurant	0.05
4	Steakhouse	0.05

----West Palm Beach----

	venue	freq
0	Bar	0.06
1	Pizza Place	0.06
2	Coffee Shop	0.04
3	French Restaurant	0.04
4	Farmers Market	0.04

----Westlake----

	venue	freq
0	Intersection	0.5
1	Baseball Field	0.5
2	Mexican Restaurant	0.0
3	Pharmacy	0.0
4	Park	0.0

```
[25]: #function to sort in decending order
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]
```

```
[26]: num_top_venues = 10

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

# create columns according to number of top venues
columns = ['MUNICIPALITIES']
```

```

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
neighborhoods_venues_sorted = pd.DataFrame(columns=columns)
neighborhoods_venues_sorted['MUNICIPALITIES'] = PBC_grouped['MUNICIPALITIES']

for ind in np.arange(PBC_grouped.shape[0]):
    neighborhoods_venues_sorted.iloc[ind, 1:] = \
    ↪return_most_common_venues(PBC_grouped.iloc[ind, :], num_top_venues)

neighborhoods_venues_sorted.head()

```

```

[26]:
MUNICIPALITIES 1st Most Common Venue \
0              Boynton Beach  American Restaurant
1  Palm Beach County Unincorporated Area  Bus Station
2              Palm Beach Gardens  Hotel
3              West Palm Beach  Bar
4              Westlake  Intersection

2nd Most Common Venue 3rd Most Common Venue 4th Most Common Venue \
0              Beach  Playground  Museum
1              Hotel  Shoe Store  Diner
2      Hotel Pool  Pharmacy  Breakfast Spot
3      Pizza Place  Farmers Market  Asian Restaurant
4  Baseball Field  Wine Bar  Gas Station

5th Most Common Venue 6th Most Common Venue 7th Most Common Venue \
0              Café  Construction & Landscaping  Greek Restaurant
1  Business Service  Rest Area  Lingerie Store
2      Chocolate Shop  Coffee Shop  Donut Shop
3              Bank  French Restaurant  Park
4              Diner  Donut Shop  Farmers Market

8th Most Common Venue 9th Most Common Venue 10th Most Common Venue
0              Gastropub  Gas Station  Concert Hall
1              Gastropub  Gas Station  Concert Hall
2              Gas Station  Greek Restaurant  Grocery Store
3  Middle Eastern Restaurant  Gastropub  Coffee Shop
4              Fountain  French Restaurant  Greek Restaurant

```

## 2.1 Cluster Neighborhoods

Run  $k$ -means to cluster the neighborhood into 5 clusters.

```
[27]: # set number of clusters
kclusters = 5

PBC_grouped_clustering = PBC_grouped.drop('MUNICIPALITIES', 1)

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

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

```
[27]: array([2, 3, 1, 4, 0], dtype=int32)
```

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

PBC_merged = PBC_df

# merge toronto_grouped with toronto_data to add latitude/longitude for each
    ↪neighborhood
PBC_merged = PBC_merged.join(neighborhoods_venues_sorted.
    ↪set_index('MUNICIPALITIES'), on='MUNICIPALITIES')

PBC_merged.head() # check the last columns!
```

```
[28]:
```

	MUNICIPALITIES	SINGLE FAMILY UNITS	SFU VALUE	\
0	Palm Beach County Unincorporated Area	543	\$191,834,714	
1	West Palm Beach	17	\$4,036,764	
2	Boynton Beach	3	\$712,370	
3	Westlake	76	\$17,596,144	
4	Palm Beach Gardens	53	\$24,789,011	

	MULTI FAMILY UNITS	MFU VALUE	TOTAL UNITS	TOTAL UNITS VALUE	Latitude	\
0	156	\$8,566,105	699	\$200,400,819	26.709723	
1	558	\$56,036,486	575	\$60,073,250	26.715300	
2	116	\$6,217,584	119	\$6,929,954	26.525349	
3	0	\$0	76	\$17,596,144	26.747000	
4	20	\$1,981,103	73	\$26,770,114	26.839600	

	Longitude	Cluster Labels	1st Most Common Venue	2nd Most Common Venue	\
0	-80.064163	3	Bus Station	Hotel	
1	-80.053400	4	Bar	Pizza Place	
2	-80.066431	2	American Restaurant	Beach	
3	-80.307700	0	Intersection	Baseball Field	
4	-80.101900	1	Hotel	Hotel Pool	



	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	\
0	Shoe Store	Diner	Business Service	
1	Farmers Market	Asian Restaurant	Bank	
2	Playground	Museum	Café	
3	Wine Bar	Gas Station	Diner	
4	Pharmacy	Breakfast Spot	Chocolate Shop	

	6th Most Common Venue	7th Most Common Venue	\
0	Rest Area	Lingerie Store	
1	French Restaurant	Park	
2	Construction & Landscaping	Greek Restaurant	
3	Donut Shop	Farmers Market	
4	Coffee Shop	Donut Shop	

	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Gastropub	Gas Station	Concert Hall
1	Middle Eastern Restaurant	Gastropub	Coffee Shop
2	Gastropub	Gas Station	Concert Hall
3	Fountain	French Restaurant	Greek Restaurant
4	Gas Station	Greek Restaurant	Grocery Store

Finally, let's visualize the resulting clusters

```
[29]: # create map
map_clusters = folium.Map(location=[latitude, longitude], zoom_start=10)

# 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(PBC_merged['Latitude'],
    ↳PBC_merged['Longitude'], PBC_merged['MUNICIPALITIES'], PBC_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
```

```
[29]: <folium.folium.Map at 0x7f42386d40b8>
```

Let's print each neighborhood along with the 20 least common venues

```
[30]: num_least_venues = 5

for hood in PBC_grouped['MUNICIPALITIES']:
    print("----"+hood+"----")
    temp = PBC_grouped[PBC_grouped['MUNICIPALITIES'] == hood].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=True).reset_index(drop=True).
    ↪head(num_least_venues))
    print('\n')
```

----Boynton Beach----

	venue	freq
0	Hotel Pool	0.0
1	Ice Cream Shop	0.0
2	Intersection	0.0
3	Irish Pub	0.0
4	Italian Restaurant	0.0

----Palm Beach County Unincorporated Area----

	venue	freq
0	American Restaurant	0.0
1	Ice Cream Shop	0.0
2	Intersection	0.0
3	Irish Pub	0.0
4	Italian Restaurant	0.0

----Palm Beach Gardens----

	venue	freq
0	Wine Bar	0.0
1	Gym	0.0
2	Gym / Fitness Center	0.0
3	Sandwich Place	0.0
4	Hobby Shop	0.0

----West Palm Beach----

	venue	freq
--	-------	------

```

0      Hotel Pool    0.0
1      Irish Pub    0.0
2  Health Food Store    0.0
3      Lingerie Store    0.0
4   Greek Restaurant    0.0

```

----Westlake----

```

      venue  freq
0  American Restaurant    0.0
1      Irish Pub    0.0
2  Italian Restaurant    0.0
3      Juice Bar    0.0
4      Lingerie Store    0.0

```

```

[159]: #function to sort in decending order
def return_least_common_venues(row, num_least_venues):
    row_categories = row.iloc[1:]
    row_categories_sorted = row_categories.sort_values(ascending=True)

    return row_categories_sorted.index.values[0:num_least_venues]

```

```

[160]: num_least_venues = 20

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

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

# create a new dataframe
neighborhoods_venues_sorted = pd.DataFrame(columns=columns)
neighborhoods_venues_sorted['MUNICIPALITIES'] = PBC_grouped['MUNICIPALITIES']

for ind in np.arange(PBC_grouped.shape[0]):
    neighborhoods_venues_sorted.iloc[ind, 1:] =
    ↪return_least_common_venues(PBC_grouped.iloc[ind, :], num_least_venues)

neighborhoods_venues_sorted.head()

```

[160]:

MUNICIPALITIES 1st Least Common Venue \			
0	Boynton Beach	Hotel Pool	
1	Palm Beach County Unincorporated Area	American Restaurant	
2	Palm Beach Gardens	Wine Bar	
3	West Palm Beach	Hotel Pool	
4	Westlake	American Restaurant	
2nd Least Common Venue 3rd Least Common Venue 4th Least Common Venue \			
0	Ice Cream Shop	Intersection	Irish Pub
1	Ice Cream Shop	Intersection	Irish Pub
2	Gym	Gym / Fitness Center	Sandwich Place
3	Irish Pub	Health Food Store	Lingerie Store
4	Irish Pub	Italian Restaurant	Juice Bar
5th Least Common Venue 6th Least Common Venue 7th Least Common Venue \			
0	Italian Restaurant	Juice Bar	Lingerie Store
1	Italian Restaurant	Juice Bar	Lounge
2	Hobby Shop	Salad Place	Train Station
3	Greek Restaurant	Museum	Gas Station
4	Lingerie Store	Lounge	Mediterranean Restaurant
8th Least Common Venue 9th Least Common Venue ... \			
0	Lounge	Mediterranean Restaurant	...
1	Mediterranean Restaurant	Mexican Restaurant	...
2	Ice Cream Shop	Intersection	...
3	Playground	Rest Area	...
4	Mexican Restaurant	Middle Eastern Restaurant	...
11th Least Common Venue 12th Least Common Venue 13th Least Common Venue \			
0	Middle Eastern Restaurant	Music Venue	Nightclub
1	Museum	Music Venue	Nightclub
2	Italian Restaurant	Sculpture Garden	Juice Bar
3	Construction & Landscaping	Hotel	Chocolate Shop
4	Music Venue	Nightclub	Park
14th Least Common Venue 15th Least Common Venue 16th Least Common Venue \			
0	Park	Train Station	Pharmacy
1	Park	Pharmacy	Pizza Place
2	Lounge	Mediterranean Restaurant	Restaurant
3	Intersection	Café	Steakhouse
4	Pharmacy	Ice Cream Shop	Pizza Place
17th Least Common Venue 18th Least Common Venue 19th Least Common Venue \			
0	Pub	Rest Area	Restaurant
1	Playground	Pub	Restaurant
2	Middle Eastern Restaurant	Museum	Music Venue
3	Baseball Field	Beach	Shoe Store

	Pub	Rest Area	Restaurant
4			
	20th Least Common Venue		
0	Road		
1	Road		
2	Nightclub		
3	Shopping Mall		
4	Road		

[5 rows x 21 columns]

Let's decide which venues are necessary and look at what Municipalities of new development they are lacking. In order to suggest these services to move into the neighborhood.

### 3 lets look at venue categories to get a better idea

```
[51]: VenueC = PBC_venues['Venue Category'].unique()
VenueC
```

```
[51]: array(['Lingerie Store', 'Hotel', 'Rest Area', 'Diner', 'Shoe Store',
        'Bus Station', 'Business Service', 'Salad Place', 'Hobby Shop',
        'Mexican Restaurant', 'Middle Eastern Restaurant', 'Bar',
        'Sushi Restaurant', 'Coffee Shop', 'Pub', 'Italian Restaurant',
        'Gastropub', 'Asian Restaurant', 'Farmers Market',
        'French Restaurant', 'Gym / Fitness Center', 'Nightclub',
        'Burrito Place', 'Wine Bar', 'American Restaurant', 'Lounge',
        'Donut Shop', 'Theater', 'Mediterranean Restaurant', 'Pizza Place',
        'Park', 'Restaurant', 'Train Station', 'Bistro', 'Brewery',
        'Juice Bar', 'Fountain', 'Sandwich Place', 'Candy Store', 'Road',
        'Ice Cream Shop', 'Bank', 'Music Venue', 'Sculpture Garden',
        'Pharmacy', 'Concert Hall', 'Breakfast Spot', 'Grocery Store',
        'Gym', 'Museum', 'Café', 'Beach', 'Construction & Landscaping',
        'Playground', 'Intersection', 'Baseball Field', 'Greek Restaurant',
        'Irish Pub', 'Health Food Store', 'Steakhouse', 'Gas Station',
        'Shipping Store', 'Shopping Mall', 'Hotel Pool', 'Chocolate Shop'],
        dtype=object)
```

**3.0.1** What we see is there is a restaurant column and then many different kinds of restaurants under their type in their own column. Also there is a Bar and a Pub and Wine Bar and Lounge and Brewery which seems the same. This means the Venue Category is confusing.

```
[139]:
```

```

new_column= np.array(['Clothing', 'Hospitality', 'Park', 'Restaurant',
↳ 'Clothing', 'Bus', 'Business Service', 'Restaurant', 'Hobby Shop',
↳ 'Restaurant', 'Restaurant', 'Bar', 'Restaurant', 'Cafe', 'Bar',
↳ 'Restaurant', 'Bar', 'Restaurant', 'Grocery', 'Restaurant', 'Gym', 'Nightclub', 'Restaurant', 'Bar',
↳ 'Restaurant', 'Train', 'Restaurant', 'Bar', 'Bar', 'Park', 'Restaurant', 'Grocery', 'Road', 'Restaur
↳ Station', 'Business Service', 'Clothing', 'Hospitality', 'Grocery'])
dataset = pd.DataFrame({'Venue Category': VenueC, 'Category': new_column},
↳ columns=['Venue Category', 'Category'])
dataset

```

```

[139]:
  Venue Category      Category
0  Lingerie Store    Clothing
1           Hotel    Hospitality
2       Rest Area         Park
3           Diner    Restaurant
4       Shoe Store    Clothing
..          ...          ...
60      Gas Station    Gas Station
61  Shipping Store  Business Service
62   Shopping Mall    Clothing
63      Hotel Pool    Hospitality
64  Chocolate Shop    Grocery

```

[65 rows x 2 columns]

This new dataframe has narrowed down the search because there were multiple of the same services in different categories

```

[140]: CategoryPBC = pd.merge(NewHousingPBC, dataset, on='Venue Category')
CategoryPBC

```

```

[140]:
  MUNICIPALITIES  SINGLE FAMILY UNITS  SFU VALUE  MULTI FAMILY UNITS  \
0   Boynton Beach                3    $712,370                116
1   Boynton Beach                3    $712,370                116
2   Boynton Beach                3    $712,370                116
3   Boynton Beach                3    $712,370                116
4   Boynton Beach                3    $712,370                116
..          ...          ...          ...          ...
102  West Palm Beach                17    $4,036,764                558
103  West Palm Beach                17    $4,036,764                558
104  West Palm Beach                17    $4,036,764                558
105      Westlake                76    $17,596,144                0
106      Westlake                76    $17,596,144                0

  MFU VALUE  TOTAL UNITS  TOTAL UNITS VALUE  Municipality Latitude  \
0   $6,217,584          119    $6,929,954          26.525349
1   $6,217,584          119    $6,929,954          26.525349

```

2	\$6,217,584	119	\$6,929,954	26.525349
3	\$6,217,584	119	\$6,929,954	26.525349
4	\$6,217,584	119	\$6,929,954	26.525349
..	...	...	...	...
102	\$56,036,486	575	\$60,073,250	26.715300
103	\$56,036,486	575	\$60,073,250	26.715300
104	\$56,036,486	575	\$60,073,250	26.715300
105	\$0	76	\$17,596,144	26.747000
106	\$0	76	\$17,596,144	26.747000

	Municipality	Longitude	Venue \
0	-80.066431	Schoolhouse Children's Museum & Learning Center	
1	-80.066431	Sailfish Cafe	
2	-80.066431	South Beach Baby	
3	-80.066431	Concrete Solutions Fl	
4	-80.066431	Mulch Park	
..	...	...	
102	-80.053400	Sandi The Sand Christmas Tree	
103	-80.053400	Ford Stage @ Sunfest	
104	-80.053400	Gym @ 610 Clematis	
105	-80.307700	Seminal Pratt & Sycamore	
106	-80.307700	Seminole Ridge Baseball Fields	

	Venue Latitude	Venue Longitude	Venue Category	Category
0	26.527577	-80.062953	Museum	Museum
1	26.526408	-80.063509	Café	Cafe
2	26.523549	-80.063925	Beach	Park
3	26.524458	-80.063360	Construction & Landscaping	Construction
4	26.526872	-80.062625	Playground	Park
..	...	...	...	...
102	26.713049	-80.049560	Sculpture Garden	Park
103	26.713149	-80.049118	Concert Hall	Theater
104	26.712944	-80.057572	Gym	Gym
105	26.746012	-80.307793	Intersection	Intersection
106	26.746715	-80.309458	Baseball Field	Park

[107 rows x 14 columns]

```
[141]: CategoryPBC.groupby('Category').count().head()
```

```
[141]:
```

	MUNICIPALITIES	SINGLE FAMILY UNITS	SFU VALUE \
Category			
Bank	3	3	3
Bar	13	13	13
Bus	2	2	2
Business Service	2	2	2
Cafe	8	8	8

	MULTI FAMILY UNITS	MFU VALUE	TOTAL UNITS \
Category			
Bank	3	3	3
Bar	13	13	13
Bus	2	2	2
Business Service	2	2	2
Cafe	8	8	8

	TOTAL UNITS VALUE	Municipality Latitude \
Category		
Bank	3	3
Bar	13	13
Bus	2	2
Business Service	2	2
Cafe	8	8

	Municipality Longitude	Venue	Venue Latitude \
Category			
Bank	3	3	3
Bar	13	13	13
Bus	2	2	2
Business Service	2	2	2
Cafe	8	8	8

	Venue Longitude	Venue Category
Category		
Bank	3	3
Bar	13	13
Bus	2	2
Business Service	2	2
Cafe	8	8

```
[142]: small_df=CategoryPBC.loc[:,['MUNICIPALITIES','Category','Venue_
→Category','Venue','TOTAL UNITS']]
small_df.head()
```

```
[142]: MUNICIPALITIES      Category      Venue Category \
0  Boynton Beach      Museum      Museum
1  Boynton Beach      Cafe      Café
2  Boynton Beach      Park      Beach
3  Boynton Beach  Construction  Construction & Landscaping
4  Boynton Beach      Park      Playground

      Venue  TOTAL UNITS
0  Schoolhouse Children's Museum & Learning Center      119
1                      Sailfish Cafe      119
```



2	South Beach Baby	119
3	Concrete Solutions Fl	119
4	Mulch Park	119

## 4 DATA FRAMES FOR EACH CITY WITH TYPES OF CATEGORY AND COUNTS OF EACH CATEGORY AND TOTAL NEW UNITS BEING BUILT IN CITY

```
[143]: small_df['MUNICIPALITIES'].unique()
```

```
[143]: array(['Boynton Beach', 'Palm Beach Gardens', 'West Palm Beach',
        'Palm Beach County Unincorporated Area', 'Westlake'], dtype=object)
```

### 4.1 BOYNTON BEACH

```
[144]: Boynton_Beach=small_df[small_df['MUNICIPALITIES']=='Boynton Beach']
        Boynton_Beach_c=Boynton_Beach['Category'].value_counts().to_frame(name='Count')
        Boynton_Beach
```

```
[144]:
```

	MUNICIPALITIES	Category	Venue Category \
0	Boynton Beach	Museum	Museum
1	Boynton Beach	Cafe	Café
2	Boynton Beach	Park	Beach
3	Boynton Beach	Construction	Construction & Landscaping
4	Boynton Beach	Park	Playground
5	Boynton Beach	Restaurant	American Restaurant

	Venue	TOTAL UNITS
0	Schoolhouse Children's Museum & Learning Center	119
1	Sailfish Cafe	119
2	South Beach Baby	119
3	Concrete Solutions Fl	119
4	Mulch Park	119
5	John G's	119

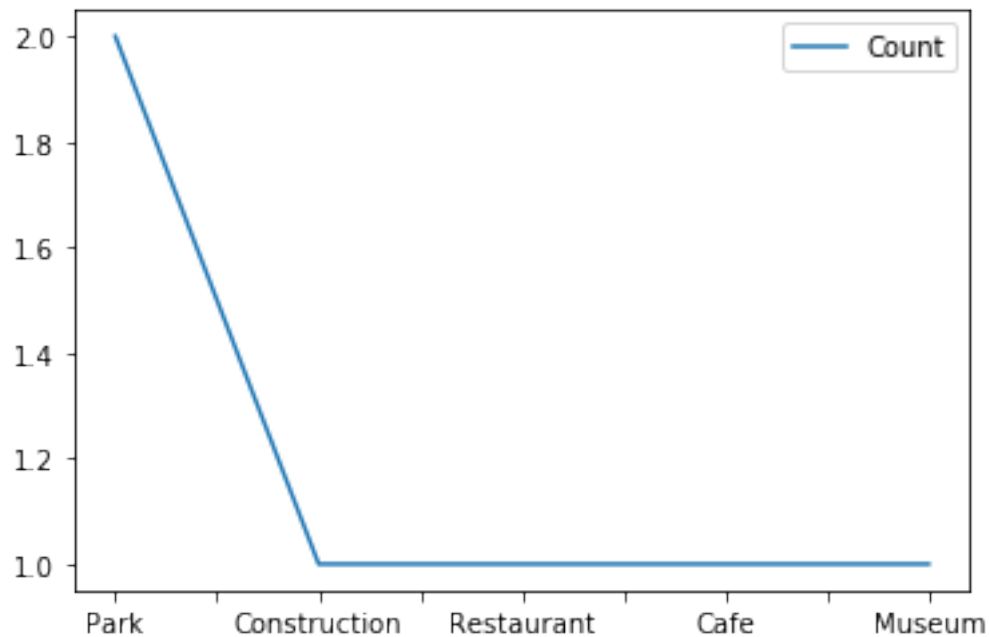
```
[156]: Boynton_Beach_c
```

```
[156]:
```

	Count
Park	2
Construction	1
Restaurant	1
Cafe	1
Museum	1

```
[158]: Boynton_Beach_c.plot()
```

[158]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7f422df04780>



## 4.2 PALM BEACH GARDENS

```
[145]: Palm_Beach_Gardens=small_df[small_df['MUNICIPALITIES']=='Palm Beach Gardens']
Palm_Beach_Gardens_c=Palm_Beach_Gardens['Category'].value_counts().
↳to_frame(name='Count')
Palm_Beach_Gardens
```

```
[145]:
```

	MUNICIPALITIES	Category	Venue Category \
6	Palm Beach Gardens	Restaurant	American Restaurant
10	Palm Beach Gardens	Hospitality	Hotel
11	Palm Beach Gardens	Hospitality	Hotel
15	Palm Beach Gardens	Clothing	Shoe Store
19	Palm Beach Gardens	Restaurant	Greek Restaurant
20	Palm Beach Gardens	Bar	Irish Pub
21	Palm Beach Gardens	Grocery	Grocery Store
23	Palm Beach Gardens	Restaurant	Mexican Restaurant
26	Palm Beach Gardens	Pharmacy	Pharmacy
28	Palm Beach Gardens	Grocery	Health Food Store
29	Palm Beach Gardens	Cafe	Coffee Shop
33	Palm Beach Gardens	Restaurant	Steakhouse
34	Palm Beach Gardens	Gas Station	Gas Station
35	Palm Beach Gardens	Cafe	Donut Shop
38	Palm Beach Gardens	Restaurant	Sushi Restaurant

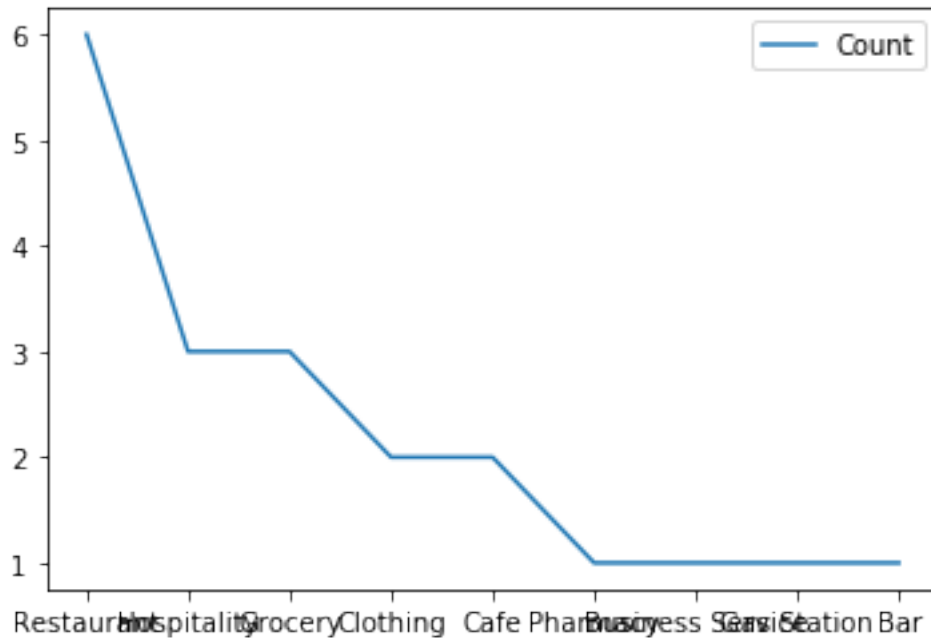
40	Palm Beach Gardens	Business Service	Shipping Store
41	Palm Beach Gardens	Restaurant	Breakfast Spot
43	Palm Beach Gardens	Clothing	Shopping Mall
44	Palm Beach Gardens	Hospitality	Hotel Pool
45	Palm Beach Gardens	Grocery	Chocolate Shop

	Venue	TOTAL UNITS
6	Cafe Chardonnay	73
10	DoubleTree by Hilton	73
11	Palm Beach Gardens Marriott	73
15	The Shoe Spa	73
19	Mr. Gyros - Mediterranean Grill	73
20	Paddy Mac's Irish Pub & Restaurant	73
21	Publix	73
23	Tacueria Taco Chula	73
26	CVS pharmacy	73
28	Mother Nature's Pantry	73
29	Starbucks	73
33	Outback Steakhouse	73
34	Shell	73
35	Dunkin'	73
38	Sushi Yama Asian Bistro	73
40	The UPS Store	73
41	Boulevard Restaurant & Gourmet	73
43	Garden Square Shops	73
44	Poolside At The Embassy Suites	73
45	Hoffman's Chocolates	73

```
[154]: Palm_Beach_Gardens_c.plot()
Palm_Beach_Gardens_c
```

```
[154]:
```

	Count
Restaurant	6
Hospitality	3
Grocery	3
Clothing	2
Cafe	2
Pharmacy	1
Business Service	1
Gas Station	1
Bar	1



### 4.3 WEST PALM BEACH

```
[147]: West_Palm_Beach=small_df[small_df['MUNICIPALITIES']=='West Palm Beach']
West_Palm_Beach_c=West_Palm_Beach['Category'].value_counts().
↳to_frame(name='Count')
West_Palm_Beach
```

```
[147]:
```

	MUNICIPALITIES	Category	Venue Category \
7	West Palm Beach	Restaurant	American Restaurant
22	West Palm Beach	Grocery	Grocery Store
24	West Palm Beach	Restaurant	Mexican Restaurant
25	West Palm Beach	Restaurant	Mexican Restaurant
27	West Palm Beach	Pharmacy	Pharmacy
..	...	...	...
100	West Palm Beach	Theater	Music Venue
101	West Palm Beach	Theater	Music Venue
102	West Palm Beach	Park	Sculpture Garden
103	West Palm Beach	Theater	Concert Hall
104	West Palm Beach	Gym	Gym

	Venue	TOTAL UNITS
7	Duffy's Sports Grill	575
22	Green Olive Market Place	575
24	Rocco's Tacos and Tequila Bar	575
25	Banko Cantina	575

27	CVS Pharmacy	575
..	...	...
100	CLUB SUNFEST	575
101	Respectable Street Cafe	575
102	Sandi The Sand Christmas Tree	575
103	Ford Stage @ Sunfest	575
104	Gym @ 610 Clematis	575

[71 rows x 5 columns]

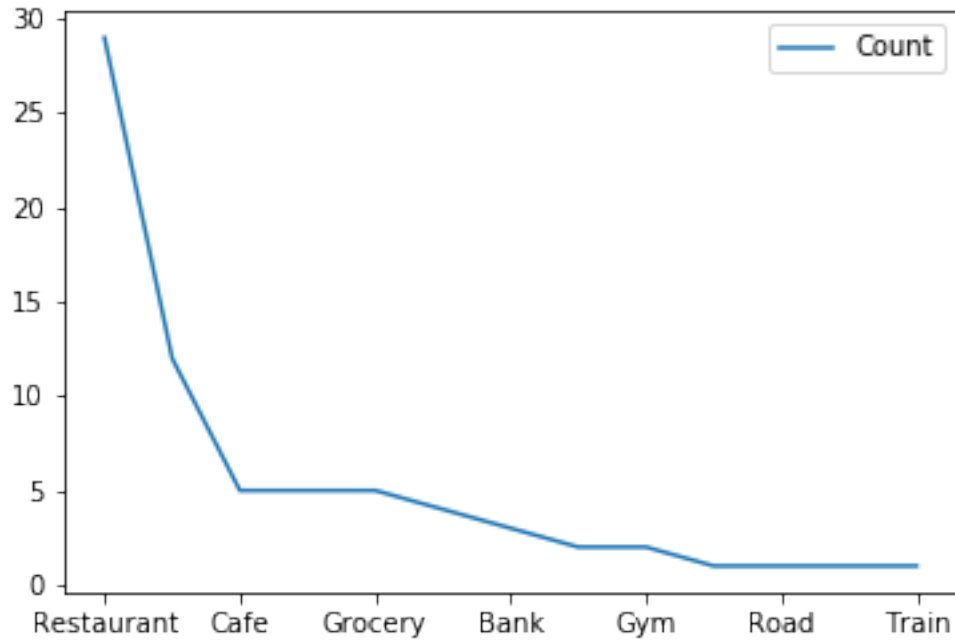
```
[148]: West_Palm_Beach_c
```

```
[148]:
```

	Count
Restaurant	29
Bar	12
Cafe	5
Park	5
Grocery	5
Theater	4
Bank	3
Nightclub	2
Gym	2
Hobby Shop	1
Road	1
Pharmacy	1
Train	1

```
[149]: West_Palm_Beach_c.plot()
```

```
[149]: <matplotlib.axes._subplots.AxesSubplot at 0x7f422e14a080>
```



#### 4.4 PALM BEACH COUNTY UNINCORPORATED AREA

```
[150]: PBC_Unincorporated=small_df[small_df['MUNICIPALITIES']=='Palm Beach County_
↳Unincorporated Area']
PBC_Unincorporated_c=PBC_Unincorporated['Category'].value_counts().
↳to_frame(name='Count')
PBC_Unincorporated
```

```
[150]:
```

	MUNICIPALITIES	Category	Venue Category \
8	Palm Beach County Unincorporated Area	Clothing	Lingerie Store
9	Palm Beach County Unincorporated Area	Hospitality	Hotel
12	Palm Beach County Unincorporated Area	Park	Rest Area
13	Palm Beach County Unincorporated Area	Restaurant	Diner
14	Palm Beach County Unincorporated Area	Clothing	Shoe Store
16	Palm Beach County Unincorporated Area	Bus	Bus Station
17	Palm Beach County Unincorporated Area	Bus	Bus Station
18	Palm Beach County Unincorporated Area	Business Service	Business Service

	Venue	TOTAL UNITS
8	Victoria's Secret PINK	699
9	West Palm Beach Marriott	699
12	Okeechobee & Parker RR Xing	699
13	Bistro Ten Zero One	699
14	Shoes For Crews LLC	699
16	Greyhound Bus Lines	699

17	West Palm Beach Intermodal Transport Center	699
18	Robert Half	699

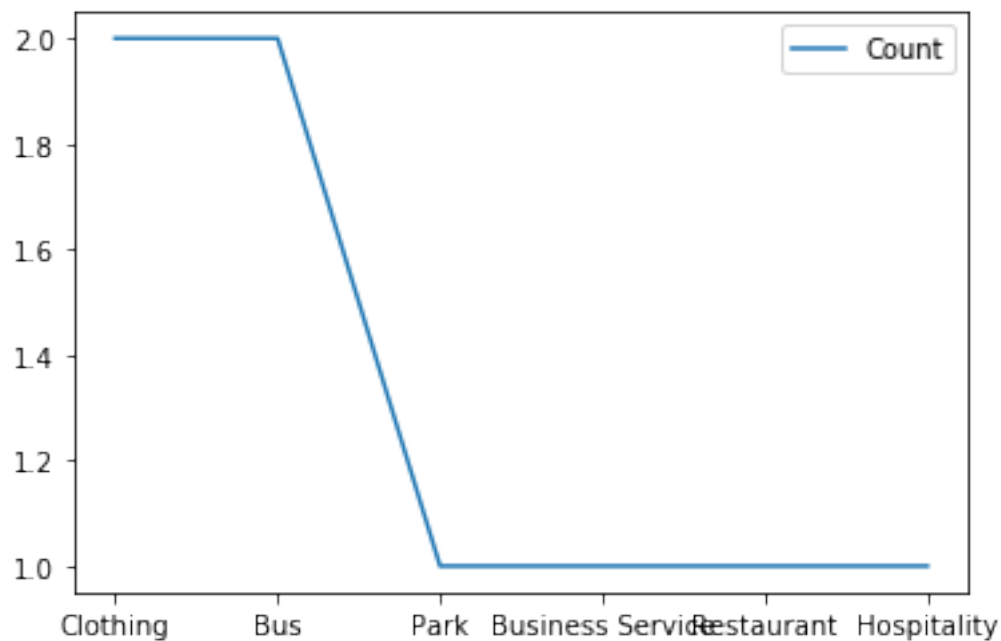
```
[151]: PBC_Unincorporated_c
```

```
[151]:
```

	Count
Clothing	2
Bus	2
Park	1
Business Service	1
Restaurant	1
Hospitality	1

```
[152]: PBC_Unincorporated_c.plot()
```

```
[152]: <matplotlib.axes._subplots.AxesSubplot at 0x7f422e0d1cc0>
```



## 4.5 WESTLAKE

```
[153]: Westlake=small_df[small_df['MUNICIPALITIES']=='Westlake']
Westlake=Westlake.drop(columns=['MUNICIPALITIES'])
Westlake_c=Westlake['Category'].value_counts().to_frame(name='Count')
Westlake_c
```

```
[153]:
```

	Count
Intersection	1
Park	1

## 5 CONCLUSION

After observing the data given regarding the venues surrounding the new housing developments in Palm Beach County where the 5 most areas are being developed in the 4th quarter of 2019. It seems that there is a need for a lot of commercial development needed. Yet since I live in Palm Beach County, I am sure that the foursquare data is incorrect and there are much more venues then what is listed. Therefore I would not make recommendations using that dataset.

by Heidi Peterson

```
[ ]:
```