

Este notebook se basa en información de target



Ahora imagina que somos parte del equipo de data science de la empresa Target, una de las tiendas con mayor presencia en Estados Unidos. El departamento de logistica acude a nosotros para saber donde le conviene poner sus almacenes, para que se optimice el gasto de gasolina, los tiempos de entrega de los productos y se disminuyan costos. Para ello, nos pasan los datos de latitud y longitud de cada una de las tiendas.

<https://www.kaggle.com/datasets/saejinmahlauheinert/target-store-locations?select=target-locations.csv>

Si quieres saber un poco más de graficas geográficas consulta el siguiente notebook

<https://colab.research.google.com/github/QuantEcon/quantecon-notebooks-datascience/blob/master/applications/maps.ipynb#scrollTo=uo2oPtSCeAOz>

```
! pip install qeds fiona geopandas xgboost gensim folium pyLDAvis descartes
```

```
Downloading click_plugins-1.1.1-py2.py3-none-any.whl (7.1 kB)  
Collecting cligj>=0.5
```

```
Downloading cligj-0.7.2-py3-none-any.whl (7.1 kB)
```

```
Requirement already satisfied: certifi in /usr/local/lib/python3.7/dist-packages
```

```
Requirement already satisfied: attrs>=17 in /usr/local/lib/python3.7/dist-packages
```

```
Requirement already satisfied: setuptools in /usr/local/lib/python3.7/dist-packages
```

```
Collecting pyproj>=2.2.0
```

Downloading pyproj-3.2.1-cp37-cp37m-manylinux2010\_x86\_64.whl (6.3 MB)

6.3 MB 36.2 MB/s

Requirement already satisfied: shapely>=1.6 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: python-dateutil>=2.7.3 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: pytz>=2017.3 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: smart-open>=1.2.1 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: branca>=0.3.0 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: Jinja2>=2.9 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: MarkupSafe>=0.23 in /usr/local/lib/python3.7/dist-packages  
 Collecting sklearn

Downloading sklearn-0.0.post1.tar.gz (3.6 kB)

Collecting funcy

Downloading funcy-1.17-py2.py3-none-any.whl (33 kB)

Requirement already satisfied: numexpr in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: joblib in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: future in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: cyclers>=0.10 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: typing-extensions in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: et-xmlfile in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: lxml in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: tenacity>=6.2.0 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: more-itertools in /usr/local/lib/python3.7/dist-packages  
 Collecting inflection>=0.3.1

Downloading inflection-0.5.1-py2.py3-none-any.whl (9.5 kB)

Requirement already satisfied: sympy in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: numba in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: llvmlite<0.40,>=0.39.0dev0 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: importlib-metadata in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: patsy>=0.5 in /usr/local/lib/python3.7/dist-packages  
 Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.7/dist-packages

Building wheels for collected packages: qeds, pyLDAvis, sklearn

Building wheel for qeds (setup.py) ... done

Created wheel for qeds: filename=qeds-0.7.0-py3-none-any.whl size=27812 sha256=8c520cc036b9730b75850b9845

Stored in directory: /root/.cache/pip/wheels/fc/8c/52/0cc036b9730b75850b9845

Building wheel for pyLDAvis (PEP 517) ... done

Created wheel for pyLDAvis: filename=pyLDAvis-3.3.1-py2.py3-none-any.whl size=1758532 sha256=c921f617bcf2667e8a68532ba2fb

Stored in directory: /root/.cache/pip/wheels/c9/21/f6/17bcf2667e8a68532ba2fb

Building wheel for sklearn (setup.py) ... done

Created wheel for sklearn: filename=sklearn-0.0.post1-py3-none-any.whl size=4256 cc/4a8bf86613aafd5b7f1b31

Successfully built qeds pyLDAvis sklearn

Installing collected packages: munch, inflection, cligj, click-plugins, sklearn  
 Successfully installed click-plugins-1.1.1 cligj-0.7.2 fiona-1.8.22 funcy-1.17

```
import pandas as pd
import numpy as np
```

```
from tqdm import tqdm
%matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
import geopandas
```

Importa la base de datos

```
url="https://raw.githubusercontent.com/marypazrf/bdd/main/target-locations.csv"
df=pd.read_csv(url)
```

Exploremos los datos.

```
df.head()
```

	name	latitude	longitude	address	phone	
0	Alabaster	33.224225	-86.804174	250 S Colonial Dr, Alabaster, AL 35007- 4657	205- 564- 2608	<a href="https://www.target.com/s">https://www.target.com/s</a>
1	Bessemer	33.334550	-86.989778	4889 Promenade Pkwy, Bessemer, AL 35022- 7305	205- 565- 3760	<a href="https://www.target.com/s">https://www.target.com/s</a>
2	Daphne	30.602875	-87.895932	1698 US Highway 98, Daphne, AL 36526-1050	251- 621- 1050	<a href="https://www.target.com/s">https://www.target.com/s</a>

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1839 entries, 0 to 1838
Data columns (total 6 columns):
#   Column      Non-Null Count  Dtype
---  -
0   name         1839 non-null   object
1   latitude     1839 non-null   float64
2   longitude    1839 non-null   float64
3   address      1839 non-null   object
4   phone        1839 non-null   object
5   website      1839 non-null   object
dtypes: float64(2), object(4)
memory usage: 86.3+ KB
```

## Definición de Latitud y Longitud

**Latitud** Es la distancia en grados, minutos y segundos que hay con respecto al paralelo principal, que es el ecuador (0°). La latitud puede ser norte y sur.

**Longitud:** Es la distancia en grados, minutos y segundos que hay con respecto al meridiano principal, que es el meridiano de Greenwich (0°). La longitud puede ser este y oeste.

```
latlong=df[["latitude","longitude"]]
```

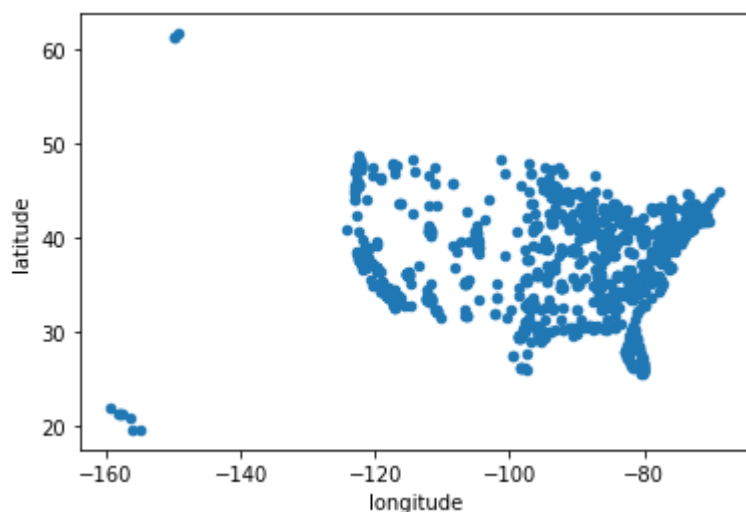
¡Visualizemos los datos!, para empezar a notar algún patron.

A simple vista pudieramos pensar que tenemos algunos datos atípicos u outliers, pero .... no es así, simplemente esta grafica no nos está dando toda la información.


```
#extrae los datos interesantes
```

```
latlong.plot.scatter( "longitude","latitude")
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f2370044590>
```



```
latlong.describe()
```

latitude longitude 

Para entender un poco más, nos auxiliaremos de una librería para graficar datos geográficos. Esto nos ayudara a tener un mejor entendimiento de ellos.

```

su      3.272299      10.100040

import geopandas as gpd
import matplotlib.pyplot as plt
import pandas as pd

from shapely.geometry import Point

%matplotlib inline
# activate plot theme
import qeds
qeds.themes.mpl_style();

df["Coordinates"] = list(zip(df.longitude, df.latitude))
df["Coordinates"] = df["Coordinates"].apply(Point)
df.head()


```

	name	latitude	longitude	address	phone	webs
0	Alabaster	33.224225	-86.804174	250 S Colonial Dr, Alabaster, AL 35007- 4657	205- 564- 2608	<a href="https://www.target.com/sl/alabaster/2">https://www.target.com/sl/alabaster/2</a>
1	Bessemer	33.334550	-86.989778	4889 Promenade Pkwy, Bessemer	205- 565-	<a href="https://www.target.com/sl/bessemer/2">https://www.target.com/sl/bessemer/2</a>

```

gdf = gpd.GeoDataFrame(df, geometry="Coordinates")
gdf.head()


```

	name	latitude	longitude	address	phone	webs
0	Alabaster	33.224225	-86.804174	250 S Colonial Dr, Alabaster, AL 35007- 4657	205- 564- 2608	<a href="https://www.target.com/sl/alabaster/2">https://www.target.com/sl/alabaster/2</a>
1	Bessemer	33.334550	-86.989778	4889 Promenade Pkwy, Bessemer	205- 565-	<a href="https://www.target.com/sl/bessemer/2">https://www.target.com/sl/bessemer/2</a>

```

#mapa


```

```
world = gpd.read_file(gpd.datasets.get_path("naturalearth_lowres"))
world = world.set_index("iso_a3")
```

```
world.head()
```

iso_a3	pop_est	continent	name	gdp_md_est	
<b>FJI</b>	920938	Oceania	Fiji	8374.0	MULTIPOLYGON (((180.
<b>TZA</b>	53950935	Africa	Tanzania	150600.0	POLYGON ((33.90371 -0
<b>ESH</b>	603253	Africa	W. Sahara	906.5	POLYGON ((-8.66559 27
<b>CAN</b>	35568668	North	Canada	1674000.0	MULTIPOLYGON (((-122

```
#graficar el mapa
world.name.unique()
```

```
array(['Fiji', 'Tanzania', 'W. Sahara', 'Canada',
      'United States of America', 'Kazakhstan', 'Uzbekistan',
      'Papua New Guinea', 'Indonesia', 'Argentina', 'Chile',
      'Dem. Rep. Congo', 'Somalia', 'Kenya', 'Sudan', 'Chad', 'Haiti',
      'Dominican Rep.', 'Russia', 'Bahamas', 'Falkland Is.', 'Norway',
      'Greenland', 'Fr. S. Antarctic Lands', 'Timor-Leste',
      'South Africa', 'Lesotho', 'Mexico', 'Uruguay', 'Brazil',
      'Bolivia', 'Peru', 'Colombia', 'Panama', 'Costa Rica', 'Nicaragua',
      'Honduras', 'El Salvador', 'Guatemala', 'Belize', 'Venezuela',
      'Guyana', 'Suriname', 'France', 'Ecuador', 'Puerto Rico',
      'Jamaica', 'Cuba', 'Zimbabwe', 'Botswana', 'Namibia', 'Senegal',
      'Mali', 'Mauritania', 'Benin', 'Niger', 'Nigeria', 'Cameroon',
      'Togo', 'Ghana', 'Côte d'Ivoire', 'Guinea', 'Guinea-Bissau',
      'Liberia', 'Sierra Leone', 'Burkina Faso', 'Central African Rep.',
      'Congo', 'Gabon', 'Eq. Guinea', 'Zambia', 'Malawi', 'Mozambique',
      'eSwatini', 'Angola', 'Burundi', 'Israel', 'Lebanon', 'Madagascar',
      'Palestine', 'Gambia', 'Tunisia', 'Algeria', 'Jordan',
      'United Arab Emirates', 'Qatar', 'Kuwait', 'Iraq', 'Oman',
      'Vanuatu', 'Cambodia', 'Thailand', 'Laos', 'Myanmar', 'Vietnam',
      'North Korea', 'South Korea', 'Mongolia', 'India', 'Bangladesh',
      'Bhutan', 'Nepal', 'Pakistan', 'Afghanistan', 'Tajikistan',
      'Kyrgyzstan', 'Turkmenistan', 'Iran', 'Syria', 'Armenia', 'Sweden',
      'Belarus', 'Ukraine', 'Poland', 'Austria', 'Hungary', 'Moldova',
      'Romania', 'Lithuania', 'Latvia', 'Estonia', 'Germany', 'Bulgaria',
      'Greece', 'Turkey', 'Albania', 'Croatia', 'Switzerland',
      'Luxembourg', 'Belgium', 'Netherlands', 'Portugal', 'Spain',
      'Ireland', 'New Caledonia', 'Solomon Is.', 'New Zealand',
      'Australia', 'Sri Lanka', 'China', 'Taiwan', 'Italy', 'Denmark',
      'United Kingdom', 'Iceland', 'Azerbaijan', 'Georgia',
      'Philippines', 'Malaysia', 'Brunei', 'Slovenia', 'Finland',
      'Slovakia', 'Czechia', 'Eritrea', 'Japan', 'Paraguay', 'Yemen',
      'Saudi Arabia', 'Antarctica', 'N. Cyprus', 'Cyprus', 'Morocco',
```

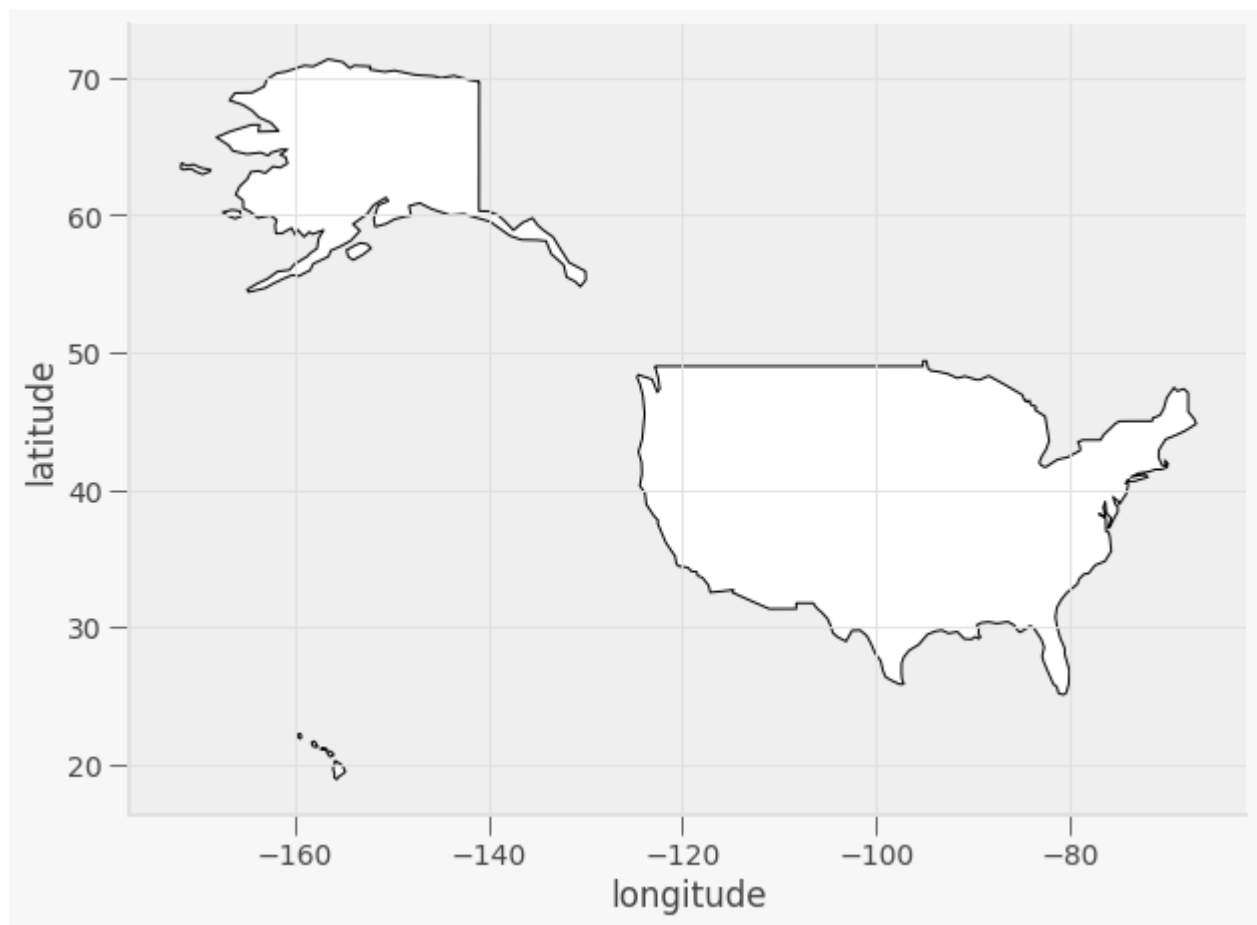
```
'Egypt', 'Libya', 'Ethiopia', 'Djibouti', 'Somaliland', 'Uganda',
'Rwanda', 'Bosnia and Herz.', 'Macedonia', 'Serbia', 'Montenegro',
'Kosovo', 'Trinidad and Tobago', 'S. Sudan'], dtype=object)
```

```
fig, gax = plt.subplots(figsize=(10,10))
```

```
# By only plotting rows in which the continent is 'South America' we only plot SA.
world.query("name == 'United States of America']").plot(ax=gax, edgecolor='black', colo
```

```
# By the way, if you haven't read the book 'longitude' by Dava Sobel, you should...
gax.set_xlabel('longitude')
gax.set_ylabel('latitude')
```

```
gax.spines['top'].set_visible(False)
gax.spines['right'].set_visible(False)
```



```
# Step 3: Plot the cities onto the map
# We mostly use the code from before --- we still want the country borders plotted --
# add a command to plot the cities
fig, gax = plt.subplots(figsize=(10,10))
```

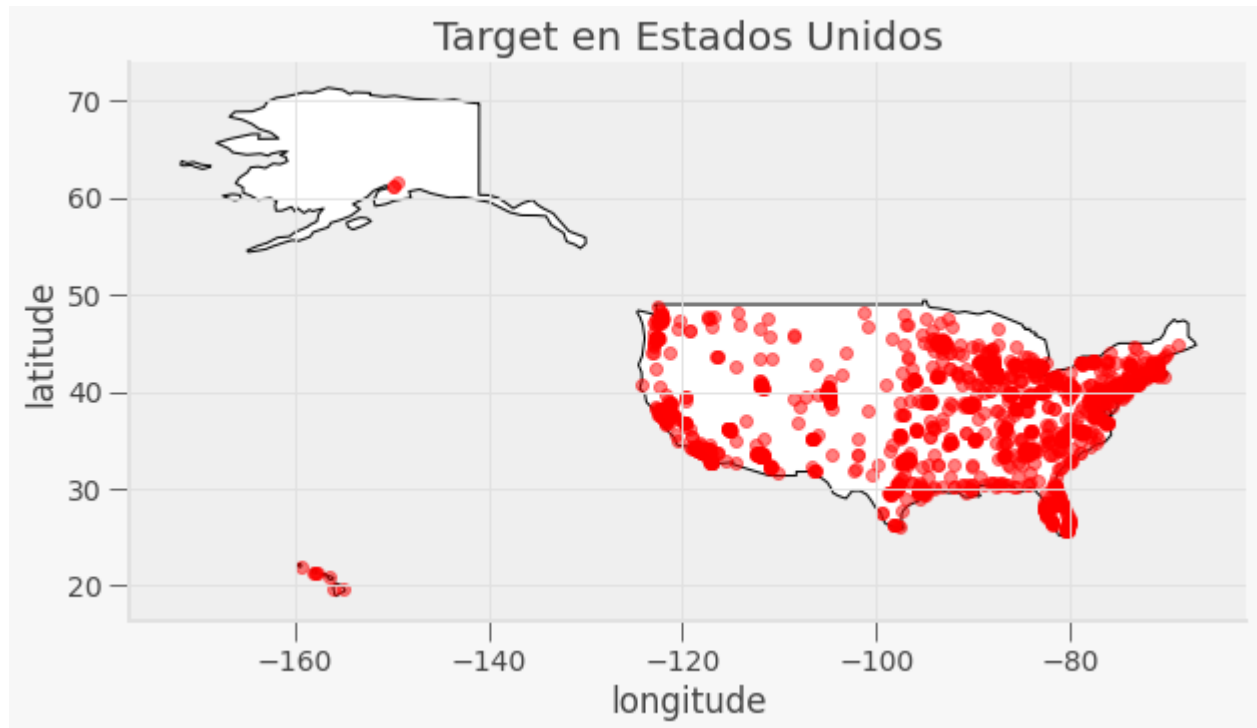
```
# By only plotting rows in which the continent is 'South America' we only plot, well,
# South America.
world.query("name == 'United States of America']").plot(ax = gax, edgecolor='black', c
```

```
# This plot the cities. It's the same syntax, but we are plotting from a different Ge
# I want the cities as pale red dots.
gdf.plot(ax=gax, color='red', alpha = 0.5)

gax.set_xlabel('longitude')
gax.set_ylabel('latitude')
gax.set_title('Target en Estados Unidos')

gax.spines['top'].set_visible(False)
gax.spines['right'].set_visible(False)

plt.show()
```



¿qué tal ahora?, tiene mayor sentido verdad, entonces los datos lejanos no eran atípicos, de aquí la importancia de ver los datos con el tipo de gráfica correcta.

Ahora sí, implementa K means a los datos de latitud y longitud :) y encuentra donde colocar los almacenes.

Nota: si te llama la atención implementar alguna otra visualización con otra librería, lo puedes hacer, no hay restricciones.

#tu código aquí

```
from sklearn.cluster import KMeans
```

```
K_clusters = range(1,10)
```

```
kmeans = [KMeans(n_clusters=i) for i in K_clusters]
```

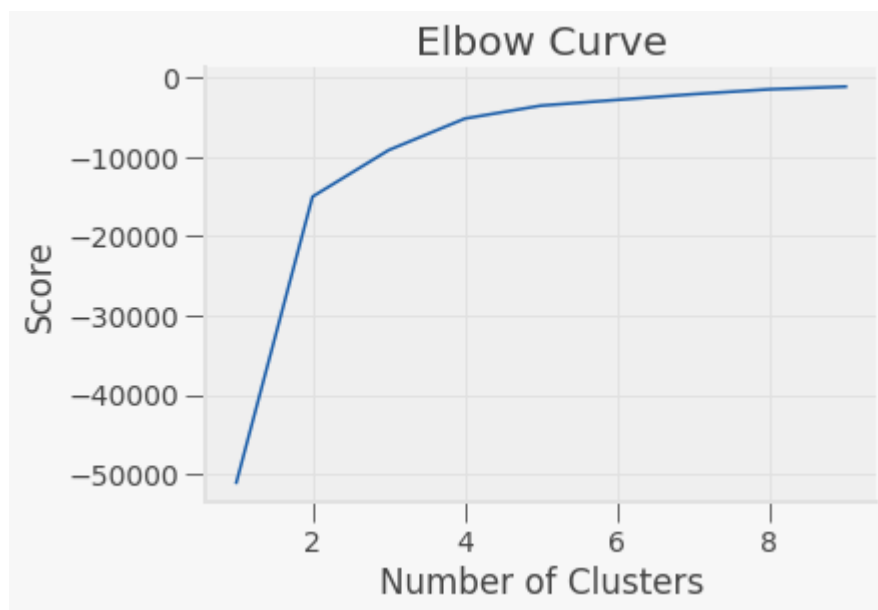


```

Y_axis = latlong[['latitude']]
X_axis = latlong[['longitude']]
score = [kmeans[i].fit(Y_axis).score(Y_axis) for i in range(len(kmeans))]

plt.plot(K_clusters, score)
plt.xlabel('Number of Clusters')
plt.ylabel('Score')
plt.title('Elbow Curve')
plt.show()

```



```

kmeans = KMeans(n_clusters = 3, init = 'k-means++')
kmeans.fit(latlong[latlong.columns[0:2]])
labels = kmeans.labels_
labels

```

```
array([2, 2, 2, ..., 1, 2, 1], dtype=int32)
```

```
X = df[["longitude", "latitude"]]
```

```

kmeans = KMeans(n_clusters=3).fit(X)
centroids = kmeans.cluster_centers_
labels = kmeans.predict(X)
C = kmeans.cluster_centers_

```

```

C_DF = pd.DataFrame(C)
C_DF["Coordinates"] = list(zip(C_DF[0], C_DF[1]))
C_DF["Coordinates"] = C_DF["Coordinates"].apply(Point)

```

```

gdf_C = gpd.GeoDataFrame(C_DF, geometry="Coordinates")
gdf_C

```

	0	1	Coordinates
0	-78.569908	37.789554	POINT (-78.56991 37.78955)
1	-93.327172	37.980063	POINT (-93.32717 37.98006)
2	-118.624473	37.487342	POINT (-118.62447 37.48734)



```
fig, gax = plt.subplots(figsize=(15,10))
```

```
world.query("name == 'United States of America']").plot(ax = gax, edgecolor='black', c
```

```
gdf.plot(ax=gax, color='red', alpha = 0.5)
```

```
gdf_C.plot(ax=gax, color='black', alpha = 1, markersize = 300)
```

```
gax.set_xlabel('longitude')
```

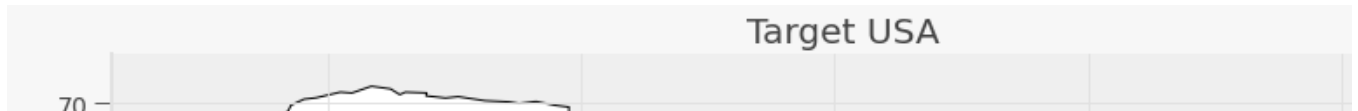
```
gax.set_ylabel('latitude')
```

```
gax.set_title('Target USA')
```

```
gax.spines['top'].set_visible(False)
```

```
gax.spines['right'].set_visible(False)
```

```
plt.show()
```



```
latlong['kmeans'] = kmeans.labels_
latlong.loc[:, 'kmeans'].value_counts()
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: <https://pandas.pydata.org/pandas-docs/stab>  
 """Entry point for launching an IPython kernel.

```
0    826
```

```
1    628
```

```
2    385
```

```
Name: kmeans, dtype: int64
```



```
gdf_C
```

	0	1	Coordinates
0	-78.569908	37.789554	POINT (-78.56991 37.78955)
1	-93.327172	37.980063	POINT (-93.32717 37.98006)
2	-118.624473	37.487342	POINT (-118.62447 37.48734)

```
from pandas.core.internals.concat import concat_arrays
```

```
Location1 = str(gdf_C[1][0]) + ", " + str(gdf_C[0][0])
```

```
print(Location1)
```

```
Location2 = str(gdf_C[1][1]) + ", " + str(gdf_C[0][1])
```

```
print(Location2)
```

```
Location3 = str(gdf_C[1][2]) + ", " + str(gdf_C[0][2])
```

```
print(Location3)
```

```
37.789554004474006, -78.56990807484885
```

```
37.98006260590112, -93.32717230430622
```

```
37.48734203064935, -118.62447331844157
```

```
from geopy.geocoders.yandex import Location
```

```
from geopy.geocoders import Nominatim
```

```
from geopy.distance import geodesic
```

```
geolocator = Nominatim(user_agent="my-application")
```

```
Locations = [Location1, Location2, Location3]
```

```
for i in Locations:
    location = geolocator.reverse(i)
    print('Almacen:', location.address)
```

```
Almacen: Langhorne Road, Totier Hills, Albemarle County, Virginia, 22946, United
Almacen: Hickory County, Missouri, United States
Almacen: Paradise Estates, Mono County, California, United States
```

```
distancia12 = str(geodesic(Location1, Location2).miles)
print("\nDistancia entre primer-segundo almacen : ", distancia12, " miles")
distancia23 = str(geodesic(Location2, Location3).miles)
print("Distancia entre segundo-tercer almacen : ", distancia23, "miles")
```

```
Distancia entre primer-segundo almacen : 805.9209470497035 miles
Distancia entre segundo-tercer almacen : 1381.7597109962394 miles
```

## Conclusiones

Encuentra las latitudes y longitudes de los almacenes, ¿qué ciudad es?, ¿a cuantas tiendas va surtir?, ¿sabes a que distancia estará? -78.569908 37.789554 POINT (-78.56991 37.78955) 1 -93.327172 37.980063 POINT (-93.32717 37.98006) 2 -118.624473 37.487342 POINT (-118.62447 37.48734)

Almacen: Langhorne Road, Totier Hills, Albemarle County, Virginia, 22946, United States Almacen: Hickory County, Missouri, United States Almacen: Paradise Estates, Mono County, California, United States

Distancia entre primer-segundo almacen : 805.9209470497035 millas Distancia entre segundo-tercer almacen : 1381.7597109962394 millas

¿Cómo elegiste el número de almacenes?, justifica tu respuesta técnicamente. Con la grafica de 'Elbow curve' se puede apreciar como la curva se va desarrollando y se pudiera agrupar las diferentes tiendas (puntos) con tan solo 3 almacenes. Mas almacenes pudieran resultar redundantes.

¿qué librerías nos pueden ayudar a graficar este tipo de datos? GeoDataFrame nos pudiera ayudar a obtener los datos de manera geografica y utilizar el matplotlib para visualizar los datos.

¿Consideras importante que se grafique en un mapa?, ¿por qué? si creo que sea importante dado que visualmente se pudiera ver mejor las agrupaciones y ver con mayor claridad los clusters.

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✓ 1s completed at 7:02 PM

