

nearest_neighbor_analysis

June 17, 2024

1 Nearest neighbor analysis

1.1 Libraries and settings

```
[ ]: # Install specific version of scipy

# Libraries
import folium
import platform
import pandas as pd
import seaborn as sns
import geopandas as gdp
import matplotlib.pyplot as plt

# Ignore warnings
import warnings
warnings.filterwarnings('ignore')

# Import functions to calculate nearest-neighbors
import nn_functions as nn
```

1.2 Prepare geodataframe of apartments data

```
[ ]: # Read apartments data
df_app = pd.read_csv('apartments_data_enriched.csv',
                    sep=',',
                    encoding='utf-8')[['web-scraper-order',
                                       'lat',
                                       'lon',
                                       'address_raw',
                                       'bfs_number',
                                       'bfs_name']]

# Convert data frame of apartments data to geodataframe
df_app_geo = gdp.GeoDataFrame(df_app,
                             geometry=gdp.points_from_xy(df_app['lon'],
                                                           df_app['lat']))
```

```
# Set Coordinate Reference System (CRS)
df_app_geo.set_crs(4326, allow_override=True)
df_app_geo.head()
```

```
[ ]:  web-scraper-order      lat      lon  \
0      1662023695-433  47.255714  8.804976
1      1662023720-634  47.254879  8.793746
2      1662023745-834  47.277386  8.800306
3      1662023701-503  47.277386  8.800306
4      1662023745-820  47.361378  8.533339

                                address_raw  bfs_number  bfs_name  \
0  Sunnenbergstrasse 15, 8633 Wolfhausen, ZH           112  Bubikon
1  Blumenbergstrasse 7, 8633 Wolfhausen, ZH           112  Bubikon
2                                8608 Bubikon, ZH           112  Bubikon
3                                8608 Bubikon, ZH           112  Bubikon
4          Lavaterstr. 63, 8002 Zürich, ZH             261   Zürich

                                geometry
0  POINT (8.80498 47.25571)
1  POINT (8.79375 47.25488)
2  POINT (8.80031 47.27739)
3  POINT (8.80031 47.27739)
4  POINT (8.53334 47.36138)
```

1.3 Prepare geodataframe of supermarkets data

```
[ ]: # Read supermarket data and select those with know brand
df_sup = pd.read_csv('supermarkets_data_enriched.csv',
                    sep=',',
                    encoding='utf-8')[['id',
                                       'lat',
                                       'lon',
                                       'brand',
                                       'bfs_number',
                                       'bfs_name']].dropna()

print(df_sup.shape)

# Convert data frame of apartments data to geodataframe
df_sup_geo = gdp.GeoDataFrame(df_sup,
                              geometry=gdp.points_from_xy(df_sup['lon'],
                                                            df_sup['lat']))

# Set Coordinate Reference System (CRS)
df_sup_geo.set_crs(4326, allow_override=True)
df_sup_geo.head()
```

```
# Subset (example)
#df_sup_geo = df_sup_geo[df_sup_geo['brand'] == 'Migros']

# Alternatively, subset of two brands (example)
df_sup_geo = df_sup_geo[df_sup_geo['brand'].isin(['Migros', 'Coop'])]

df_sup_geo.head()
```

(2009, 6)

```
[ ]:      id      lat      lon  brand  bfs_number  bfs_name  \
4   36726161  47.226191  8.980329  Migros         3339   Uznach
5   39768209  47.225069  8.969981   Coop         3339   Uznach
7   39947904  47.376732  8.542161   Coop          261   Zürich
8   48932835  47.375020  8.522895  Migros          261   Zürich
10  79977755  47.340070  8.530546   Coop          261   Zürich

      geometry
4  POINT (8.98033 47.22619)
5  POINT (8.96998 47.22507)
7  POINT (8.54216 47.37673)
8  POINT (8.52290 47.37502)
10 POINT (8.53055 47.34007)
```

1.4 Identify closest supermarkets per apartment and calculate its distance

```
[ ]: # Closest supermarket of each apartment
closest_supermarkets = nn.nearest_neighbor(df_app_geo,
                                           df_sup_geo,
                                           return_dist=True)

print(len(closest_supermarkets), '==', len(df_app_geo))

# Rename the geometry of closest stops gdf so that we can easily identify it
closest_supermarkets = closest_supermarkets.rename(columns={'geometry': ↵
    ↵ 'closest_sup_geom'})
closest_supermarkets.head()
```

870 == 870

```
[ ]:      id      lat      lon  brand  bfs_number      bfs_name  \
0   1362066985  47.229393  8.821159  Migros         3340  Rapperswil-Jona
1    956494681  47.253231  8.773446   Coop          153   Hombrechtikon
2    956494681  47.253231  8.773446   Coop          153   Hombrechtikon
3    956494681  47.253231  8.773446   Coop          153   Hombrechtikon
4    262400822  47.364072  8.530945  Migros          261      Zürich
```

		closest_sup_geom	distance
0	POINT	(8.82116 47.22939)	3406.381465
1	POINT	(8.77345 47.25323)	2264.541956
2	POINT	(8.77345 47.25323)	3995.788117
3	POINT	(8.77345 47.25323)	3995.788117
4	POINT	(8.53094 47.36407)	398.327277

1.5 Merge closest supermarkets to apartments

```
[ ]: # Merge supermarkets to apartments
result = pd.merge(closest_supermarkets,
                  df_app_geo,
                  left_index=True,
                  right_index=True)[['web-scraper-order',
                                     'address_raw',
                                     'lat_y',
                                     'lon_y',
                                     'id',
                                     'brand',
                                     'geometry',
                                     'closest_sup_geom',
                                     'distance']]

# Rename columns
results = result.rename(columns={'lat_y': 'lat',
                                'lon_y': 'lon'},
                        inplace = True)

result.head()
```

```
[ ]: web-scraper-order      address_raw      lat \
0      1662023695-433  Sonnenbergstrasse 15, 8633 Wolfhausen, ZH  47.255714
1      1662023720-634   Blumenbergstrasse 7, 8633 Wolfhausen, ZH  47.254879
2      1662023745-834                        8608 Bubikon, ZH  47.277386
3      1662023701-503                        8608 Bubikon, ZH  47.277386
4      1662023745-820                Lavaterstr. 63, 8002 Zürich, ZH  47.361378
```

	lon	id	brand	geometry \
0	8.804976	338156406	Volg	POINT (8.80498 47.25571)
1	8.793746	956494681	Coop	POINT (8.79375 47.25488)
2	8.800306	338156406	Volg	POINT (8.80031 47.27739)
3	8.800306	338156406	Volg	POINT (8.80031 47.27739)
4	8.533339	262400822	Migros	POINT (8.53334 47.36138)

		closest_sup_geom	distance
0	POINT	(8.82088 47.26967)	2340.709105
1	POINT	(8.77345 47.25323)	2264.541956
2	POINT	(8.82088 47.26967)	2439.597524

```
3 POINT (8.82088 47.26967) 2439.597524
4 POINT (8.53094 47.36407) 398.327277
```

1.6 Summary statistics of distance to closest supermarkets

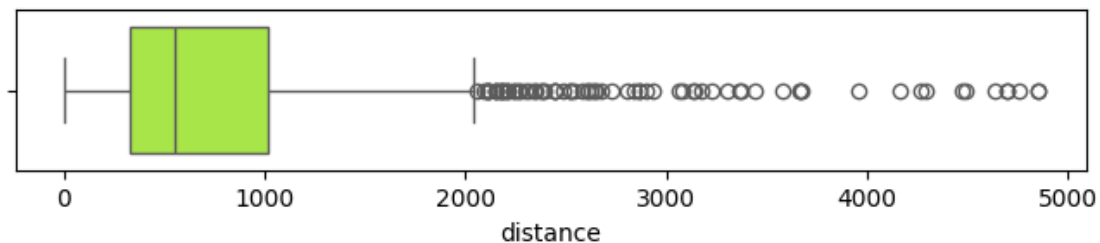
```
[ ]: result['distance'].describe()
```

```
[ ]: count      870.000000
      mean       818.831571
      std        810.177551
      min         2.198898
      25%        324.456254
      50%        550.585590
      75%       1016.089056
      max       4847.253138
      Name: distance, dtype: float64
```

1.7 Boxplot of distance to closest supermarkets

```
[ ]: plt.figure(figsize=(8,1.2))
      plt.ticklabel_format(style='plain')
      sns.boxplot(x=result['distance'],
                  color="greenyellow")
```

```
[ ]: <Axes: xlabel='distance'>
```



1.8 Plotting map with apartments and nearest supermarkets

```
[ ]: # Polygonmap als .json-File (WGS84)
      polys = gdp.read_file("GEN_A4_GEMEINDEN_2019_epsg4326.json")

      # Marker symbols
      url_01 = 'https://raw.githubusercontent.com/pointhi/leaflet-color-markers/
               ↪master/img/marker-icon-blue.png'
      url_02 = 'https://raw.githubusercontent.com/pointhi/leaflet-color-markers/
               ↪master/img/marker-icon-gold.png'
```

```

# Initialisierung der Map
m = folium.Map(location=[47.44, 8.65],
                # tiles='Stamen Toner',
                zoom_start=11)

# Plot Polygonmap of municipalities
folium.Choropleth(
    geo_data=polys,
    name='polys',
    fill_color='transparent',
    line_color='darkred').add_to(m)

# Add lat/lon of apartments
for i in range(0, len(result)):
    folium.Marker(location=(result.iloc[i]['lat'],
                           result.iloc[i]['lon']),
                  popup=result.iloc[i]['address_raw'],
                  icon=folium.features.CustomIcon(url_01, icon_size=(14, 23))).
        ↪add_to(m)

# Add lat/lon of apartments
for i in range(0, len(closest_supermarkets)):
    folium.Marker(location=(closest_supermarkets.iloc[i]['lat'],
                           closest_supermarkets.iloc[i]['lon']),
                  popup=closest_supermarkets.iloc[i]['brand'],
                  icon=folium.features.CustomIcon(url_02, icon_size=(14, 23))).
        ↪add_to(m)

# Layer control
folium.LayerControl().add_to(m)

# Plot map
m

```

```
[ ]: <folium.folium.Map at 0x1fdf0b36f00>
```

1.9 Save data to file

```

[ ]: result.to_csv('apartments_data_with_supermarkets.csv',
                  sep=",",
                  encoding='utf-8',
                  index=False)

```