OpenStreetMap Project Data Wrangling with MongoDB Bhavya Garg

Map Area: DC-Baltimore

INTRODUCTION

I chose this particular area because it is my neighbor city. This domain knowledge was very helpful as I could clean the data with some context instead of doing guesswork or doing extensive research on another city. I chose the DC-Baltimore, instead of the Richmond city from where i belong to because of the limited data that has been populated so far for Richmond on OpenStreetMaps and I needed to analyze a dataset larger than 50 MB for this project. Below is the list of .py files which shows my full work of this project.

Section 1: Problems Encountered in the Map

Unexpected Tags

mapparser.py was used to count occurrences of each tag, with a result:

member: 4765nd: 488617node: 418014

osm:1

relation: 335tag: 251958way: 40621

Then in Users.py, I find out the number of unique users who have contributed to the map in this particular area!

Street type abbreviations standardized

I changed many street types to a non-abbreviated form after discovering these issues in the audit (before importing into MongoDB). For example, below are the 3 types of street which I standardized by adding them to the mapping dictionary in the audit file (Audit.py).

```
'St': 'Street', 'St.': 'Street', 'ST': 'Street'
```

Other than Street below are the list of changes i made to the variable mapping:

- "Ave": "Avenue",
- "NW": "North West",
- "E": "East",
- "E.":"East",

```
"W.":"West",
"N.":"North",
"S.":"South",
"West":"West",
"Northeast": "North East",
"SE": "South East",
"Blvd": "Boulevard",
"Blvd.":"Boulevard",
"CIrcle": "Circle",
"Hwy": "Highway",
"Rd.": "Road",
"St.":"Street",
"St": "Street",
"ST": "Street",
"Rdt":"Road",
"Ln": "Lane",
"SW": "South West",
"Ct": "Court",
"Cir": "Circle",
"AVE": "Avenue"
```

Postal Codes

Postal code strings posed a different sort of problem, forcing a decision to strip all leading and trailing characters before and after the main 5digit zip code. This effectively dropped all leading state characters and following a hyphen . This 5digit restriction allows for more consistent queries. Postal Codes have been audited and cleaned in Audit.py and Data.py

Section 2: Data Overview

This section contains basic statistics about the dataset and the MongoDB queries used to gather them. The queries are included in query.py.

File sizes:

```
> collection.dataSize()
106068294
```

• sample.osm: 90.3 MB

sample.osm.json: 97.6 MB

Number of documents:

```
> collection.count()
449119
```

Number of nodes and ways:

```
> collection.find({'type':'node'}).count()
418014
> collection.find({'type':'way'}).count()
31105
```

Number of unique users:

```
collection.distinct("created.user").length
1176
```

Top 10 contributing user:

```
> collection.aggregate([{"$group":{ "_id":"$created.user",
"count":{"$sum":1}}},{"$sort":{"count":-1}},{"$limit":10}]).pretty()

{ "_id" : "woodpeck_fixbot", "count" : 94731 }
 { "_id" : "asciiphil", "count" : 86544 }
 { "_id" : "aude", "count" : 64797 }
 { "_id" : "mdroads", "count" : 15728 }
 { "_id" : "kriscarle", "count" : 15031 }
 { "_id" : "wonderchook", "count" : 13585 }
 { "_id" : "JoshD", "count" : 10824 }
 { "_id" : "RJCorazza", "count" : 10454 }
 { "_id" : "Evanator", "count" : 9666 }
 { "_id" : "westendguy", "count" : 9204 }
```

Number of users contributing only once:

5 Most common street address:

```
> collection.aggregate([{
                    '$match': {
                        'address.street': {
                             '$exists': 1
. . . . . .
                     '$group': {
                        ' id': '$address.street',
                        'count': {
                            '$sum': 1
                     '$sort': {
                        'count': -1
. . . . . .
                }, {
                     '$limit': 5
                }])
{ "_id" : "Connecticut Avenue North West", "count" : 45 }
 "_id" : "13th Street North West", "count" : 45 }
{ "_id" : "Q Street North West", "count" : 41 }
{ "_id" : "16th Street North West", "count" : 41 }
{ "_id" : "Georgia Avenue North West", "count" : 41 }
```

Sort cities by count, descending

```
'count': {
'$sum': 1
                                        '$sort': {
                                               'count': -1
}])
{ "_id" : "Leesburg", "count" : 137 }
{ "_id" : "Sterling", "count" : 15 }
{ "_id" : "Frederick", "count" : 13 }
{ "_id" : "Alexandria", "count" : 9 }
{ "_id" : "Baltimore", "count" : 9 }
  "_id" : "Towson", "count" : 8 }
"_id" : "Silver Spring", "count" : 7 }
  "_id" : "Pikesville", "count" : 5 }

"_id" : "Arlington", "count" : 5 }

"_id" : "Columbia", "count" : 5 }
   "_id" : "Cockeysville", "count" : 5 }
  "_id" : "Churchton", "count" : 4 }
"_id" : "Washington", "count" : 4 }
  "_id" : "Mertford Street", "count" : 4 }
  __id" : "Deale", "count" : 3 }
  "_id" : "Springfield", "count" : 3 }
"_id" : "Westminster", "count" : 3 }
{ " id" : "Vienna", "count" : 3 }
{ "_id" : "Kensington", "count" : 3 }
{ "_id" : "DC", "count" : 2 }
```

Nodes without addresses:

Sort postcodes by count, descending

```
> collection.aggregate([{
                           $match': {
                               'address.postcode': {
                                   '$exists': 1
                          '$group': {
                               'id': '$address.postcode',
                               'count': {
                                   '$sum': 1
                          '$sort': {
                               'count': -1
                     }])
 " id" : "20176",
                    "count" : 106 }
  "id": "21043", "count": 42 }
 _id" : "20175",
                    "count" : 32 }
   id" : "21093",
                    "count" : 28 }
  _id" : "21228", "count" : 21 }
  _id" : "21204", "count" : 17 }
 "_id" : "20164", "count" : 14 }
"_id" : "21703", "count" : 13 }
   id" : "21045",
                    "count" : 11 }
  " id" : "21208",
                    "count" : 9 }
  __id" : "20878", "count" : 8 }
 __id" : "21042",
                    "count" : 8 }
 "id": "21044",
                    "count" : 7 }
 _
"id" : "22314",
                    "count" : 6 }
  __id" : "21286", "count" : 6 }
   _id" : "21030", "count" : 6 }
id" : "21046", "count" : 6 }
 "_id" : "21046",
   _id" : "21061", "count" : 5 }
 __id" : "20701", "count" : 5 }
  "_id" : "21224", "count" : 4 }
```

Conclusion

There are still several opportunities for cleaning and validation that I left unexplored. Of note, the data set is populated only from one source: OpenStreetMaps. While this crowd sourced repository pulls from multiple sources, some of data is potentially outdated. It would have been an interesting exercise to pull missing information from Google Map API, since every node has latitude-longitude coordinate. Also Depending on the location we are interested in, Google Maps can provide additional information we might find useful, such as the location of a business with the general rating and reviews. But despite having the benefits of using Google Maps the information provided by the Google Maps website may be inaccurate or out of date. Walking and cycling directions, in particular, may suggest routes that are missing sidewalks, pedestrian paths or cycling paths. After fairly thoroughly exploring the data for the DC areas and surrounding towns, it is apparent that recreation-lovers have done a good job mapping the key recreation facilities. However, much of the rest of the map remains fairly incomplete -- very few

businesses, restaurants, and other useful amenities are mapped. The buildings are well-mapped, so someone just needs to add the nodes for these other amenities. The benefits in implementing the improvement is now we have a Street type abbreviations in a standardized manner. Inspite of deficiencies pointed earlier, the dataset was successfully cleaned and could be used as a standard for future contributors. Despite the efforts of some hard working users, the mapping community for the area is still very weak.