

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 : 4765
- nd : 488617
- node : 418014
- osm : 1
- relation : 335
- tag : 251958
- way : 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",
- "Bld": "Boulevard",
- "Bld.": "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"

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:

```
> coll.dataSize()
329734167
```

- sample.osm: 90.3 MB
- sample.osm.json: 100 MB

Number of documents:

```
> coll.count()
1375905
```

Number of nodes and ways:

```
> coll.find({'type': 'node'}).count()
1254042
```

```
> coll.find({'type':'way'}).count()
121863
```

Number of unique users:

```
coll.distinct("created.user").length
1194
```

Top 10 contributing user:

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

{ "_id" : "woodpeck_fixbot", "count" : 284193 }
{ "_id" : "asciiphil", "count" : 262362 }
{ "_id" : "aude", "count" : 194694 }
{ "_id" : "kriscarle", "count" : 50322 }
{ "_id" : "mdroads", "count" : 47718 }
{ "_id" : "wonderchook", "count" : 40755 }
{ "_id" : "RJCorazza", "count" : 33381 }
{ "_id" : "JoshD", "count" : 32724 }
{ "_id" : "Evanator", "count" : 30183 }
{ "_id" : "Your Village Maps", "count" : 29133 }
```

Number of users contributing only once:

```
> coll.aggregate([
... ..    {'$group': {
... ..        '_id': '$created.user',
... ..        'count': {
... ..            '$sum': 1
... ..        }
... ..    }, {
... ..        {'$group': {
... ..            '_id': '$count',
... ..            'num_users': {
... ..                '$sum': 1
... ..            }
... ..        }, {
... ..            {'$sort': {
... ..                '_id': 1
... ..            }
... ..        }, {
... ..            '$limit': 1
... ..        }
... ..    })
{ "_id" : 3, "num_users" : 327 }
```

Most common street address:

```
> coll.aggregate([
... ..    '$match': {
... ..    'address.street': {
... ..    '$exists': 1
... ..    }
... ..    }, {
... ..    '$group': {
... ..    '_id': '$address.street',
... ..    'count': {
... ..    '$sum': 1
... ..    }
... ..    }, {
... ..    '$sort': {
... ..    'count': -1
... ..    }
... ..    }, {
... ..    '$limit': 1
... ..    })
{ "_id" : "13th Street NW", "count" : 135 }
```

Sort cities by count, descending

```
> coll.aggregate([
... ..    '$match': {
... ..    'address.city': {
... ..    '$exists': 1
... ..    }
... ..    }, {
... ..    '$group': {
... ..    '_id': '$address.city',
... ..    'count': {
... ..    '$sum': 1
... ..    }
... ..    }, {
... ..    '$sort': {
... ..    'count': -1
... ..    }
... ..    })
{ "_id" : "Leesburg", "count" : 411 }
{ "_id" : "UNIVERSITY PARK", "count" : 186 }
{ "_id" : "Alexandria", "count" : 108 }
{ "_id" : "Baltimore", "count" : 81 }
{ "_id" : "Sterling", "count" : 72 }
{ "_id" : "Towson", "count" : 51 }
{ "_id" : "Frederick", "count" : 51 }
{ "_id" : "Herndon", "count" : 42 }
```

```
{ "_id" : "Arlington", "count" : 24 }
{ "_id" : "Silver Spring", "count" : 21 }
{ "_id" : "Cockeysville", "count" : 21 }
{ "_id" : "Falls Church", "count" : 18 }
{ "_id" : "Washington", "count" : 18 }
{ "_id" : "Pikesville", "count" : 15 }
{ "_id" : "Columbia", "count" : 15 }
{ "_id" : "Randallstown", "count" : 15 }
{ "_id" : "Deale", "count" : 12 }
{ "_id" : "Churchton", "count" : 12 }
{ "_id" : "Washington, DC", "count" : 12 }
{ "_id" : "Kensington", "count" : 12 }
```

Nodes without addresses:

```
> coll.aggregate([
... ..    '$match': {
... ..        'type': 'node',
... ..        'address': {
... ..            '$exists': 0
... ..        }
... ..    }, {
... ..        '$group': {
... ..            '_id': 'Nodes without addresses',
... ..            'count': {
... ..                '$sum': 1
... ..            }
... ..        }
... ..    })
{ "_id" : "Nodes without addresses", "count" : 1249698 }
```

Sort postcodes by count, descending

```
> coll.aggregate([
... ..    '$match': {
... ..        'address.postcode': {
... ..            '$exists': 1
... ..        }
... ..    }, {
... ..        '$group': {
... ..            '_id': '$address.postcode',
... ..            'count': {
... ..                '$sum': 1
... ..            }
... ..        }
... ..    }, {
... ..        '$sort': {
... ..            'count': -1
... ..        }
... ..    })
```

```
{ "_id" : "20176", "count" : 321 }
{ "_id" : "21043", "count" : 279 }
{ "_id" : "20782", "count" : 186 }
{ "_id" : "21228", "count" : 138 }
{ "_id" : "21093", "count" : 96 }
{ "_id" : "22314", "count" : 96 }
{ "_id" : "20175", "count" : 96 }
{ "_id" : "21042", "count" : 90 }
{ "_id" : "21204", "count" : 66 }
{ "_id" : "21286", "count" : 54 }
{ "_id" : "21703", "count" : 51 }
{ "_id" : "20164", "count" : 42 }
{ "_id" : "20171", "count" : 42 }
{ "_id" : "21227", "count" : 36 }
{ "_id" : "21208", "count" : 36 }
{ "_id" : "21045", "count" : 33 }
{ "_id" : "21224", "count" : 30 }
{ "_id" : "20147", "count" : 30 }
{ "_id" : "21146", "count" : 30 }
{ "_id" : "21244", "count" : 27 }
```

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 crowdsourced repository pulls from multiple sources, some of data is potentially outdated.