OpenStreetMap Data Case Study - Project 3

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Map Area: I selected Columbus, Ohio, USA as it is the city I've spent the most time in.

• Location: Columbus, Ohio

OpenStreetMap URL

Data Audit

Unique Tags

The XML file utilizes many tags to structure the data. Using mapparser.py I counted the number of each unique tag respectively from columbusOH.osm. The code for this is mapparser.py taken from the class case study.

'node': 1502751
'member': 38206,
'nd': 1837440,
'tag': 687442,
'bounds': 1, S
'note': 526,
'meta': 91356,
'relation': 77866
'way': 177854

Patterns

• 'osm' : 1

This set of pattern checks was run on the entire osmfile vs the sample that's in the root folder and utilizes regular expressions. tags.py contains the necessary code to count these 4 categories.

- "lower": 429525 Tags that only contain low ercase letters and pass the validity checks.
- "lower_colon" : 244778 Tags that are otherwise valid, yet have colons in their name.
- "problemchars" : 0 Tags with problematic characters such as "=", "+", "&", ",", "?" and more.
- \bullet "other" : 13139 Any other tags that don't fit in the 3 prior categories

Problems In Data

The most significant issue that had to be addressed was the inconsistency in the street names and their abbreviations. In order to correct these the following functions were utilized within the audit.py file:

- $\bullet \quad {\tt audit_street_type} \ : \textbf{Determines if the street name is within the list of expected names}$
- $\bullet \quad \mathsf{street_name} \ : \mathsf{Tests} \ \mathsf{w} \ \mathsf{hether} \ \mathsf{the} \ \mathsf{'k'} \ \mathsf{attribute} \ \mathsf{matches} \ \mathsf{the} \ \mathsf{key} \ \mathsf{for} \ \mathsf{street} \ \mathsf{data} \ (\ \mathsf{addr:street} \)$
- audit : Returns a dictionary of key, value pairs which meet the criteria in the preceding functions
- update_name : Actually does the update on the street name

Data Overview

File Sizes

columbusOH.osm: 324 MB
nodes.csv: 124MB
nodes_tags.csv: 3.37 MB
ways.csv: 10.5 MB
ways_nodes.csv: 43.7 MB
ways_tags.csv: 20.2 MB
cbus.db: 171 MB

${\bf Number\ of\ Nodes:}$

sqlite> SELECT COUNT(*) FROM NODES

Output: 1502751

Number of Ways:

```
sqlite> SELECT COUNT(*) FROM NODES
```

Output: 177854

A Count by type of the top 15 Node Tags

```
sqlite> SELECT DISTINCT TYPE, COUNT(ID) as TYPE_COUNT
FROM NODES_TAGS
GROUP BY TYPE
ORDER BY TYPE_COUNT DESC
LIMIT 15;
```

Output:

```
regular
             l 76620
addr
              7879
gnis
              4409
species
              | 1396
fire_hydrant
             572
traffic_signals | 404
brand
contact
              | 121
payment
              | 79
              77
tower
name
              | 50
xmas
historic
              42
service
              | 39
surveillance
             | 38
```

Number of Unique users:

```
sqlite> SELECT COUNT(DISTINCT(u.uid))
FROM (SELECT uid FROM nodes
UNION ALL
SELECT uid from ways) u;
```

Output: 1151

Top Contributors:

Output:

```
        woodpeck_fixbot
        | 211799

        doktorpixel14
        | 167004

        Anonononon
        | 157726

        Nimbalo
        | 150409

        MerlinPendragon
        | 108529

        duck57
        | 88834

        AndrewSP37
        | 87877

        Vid the Kid
        69057

        kbzimmer
        | 61976

        St-Motel
        | 53533
```

Popular Restaurants by Cuisine

```
sqlite> SELECT NODES_TAGS.VALUE, COUNT(*) AS NUM
FROM NODES_TAGS
JOIN (SELECT DISTINCT(ID) FROM NODES_TAGS
WHERE VALUE="restaurant") r on nodes_tags.id=r.id
WHERE NODES_TAGS.KEY = 'cuisine'
```

```
GROUP BY nodes_tags.value
ORDER BY NUM DESC
LIMIT 15;
```

Output:

```
| 33
pizza
chinese
                 | 22
mexican
                 | 16
sandwich
                 1 14
italian
                 10
asian
ice_cream
                 18
                 | 6
indian
                 | 5
greek
japanese
sushi
                 | 4
burger
chicken;american
                 | 3
barbecue
```

Greatest number of Worship Centers by Religion:

```
sqlite> SELECT NODES_TAGS.VALUE, COUNT(*) AS NUM
FROM NODES_TAGS
JOIN (SELECT DISTINCT(ID) FROM NODES_TAGS
WHERE VALUE = "place_of_worship") a
ON NODES_TAGS.ID = A.ID
WHERE NODES_TAGS.KEY = "religion"
GROUP BY NODES_TAGS.VALUE
ORDER BY NUM DESC
LIMIT 3;
```

Output:

```
christian | 536
muslim | 2
jewish | 1
```

Conclusion

Considering the amount of data, and how it is all manually added it is surprisingly clean. How ever that doesn't mean it IS clean. Each layer reveals more issues that would need to be addressed such as returning lists of counties. Many times there are more than one in the result. Due to the size of the data, and the non-standard tags it can be very easy to have duplication based on different colloquial names.

Suggestion: Set a reasonable standard on tags for usage so that things like restaurant:cuisine = american and chicken;american doesn't happen. I'm not quite sure w hat "chicken;america" means.