Wrangle OpenStreetMap Data

August 9, 2017

1 Wrangle OpenStreetMap Data

For this map data wrangling project, I selected my home state of Rhode Island. It is about the size of some larger urban areas, so I figured it might pose an interesting choice! The map data I downloaded from Geofabrik, which already has a pre made Rhode Island openstreetmap data file. The area extends into the ocean to include Block Island, an offshore town in the state.

1.1 Data Investigation and Auditing

Several issues were identified in the data: *Inconsistent and abbreviated street names (Point Judith Rd, Cunningham Sq) * Incorrect or inconsistent zip codes (02835, 02903-2996, 029212) * Street names embedded in Tiger GPS blocks in ways

1.1.1 Cleaning Street Data

First, like the case study, I looked into street abbreviations and found that there were many nodes with inconsistencies. Fixing these abbreviations will help standardize the data and make for easier analysis of addresses if needed. The update_street_name code was used in the conversion to csv to fix the street names.

```
def update_street_name(name, mapping):
    """
    Takes in an abnormal street name and returns a corrected one
    """

# Extract the street type and correct if possible
    street_type = re.search(street_type_re, name).group()
    if street_type in mapping:
        name = name.replace(street_type, mapping[street_type], 1)
    return name
```

This code would take something like Point Judith Rd and replace it with Point Judith Road.

1.1.2 Update Postcodes

Another important facet of the address is the post code - auditing this is not particularly difficult and can easily find some mistakes. Upon searching through the postal codes in the OSM database,

it was clear that most had the standard five digit form. However, some included the final four digits after the hyphen and a few were the wrong length entirely, probably due to human input.

The function update_postcodes is used during csv conversion to check for inconsistent postal codes. It will omit records that are likely a result of typos, and only return the first five digits of a postcode found with the hyphen format.

```
def update_postcodes(code):
    """
    Takes a postcode and updates it to a valid 5 digit postcode
    """
    # If the postcode is the proper length, return it
    if len(code) == 5:
        return code
    # Otherwise, if the postcode has a dash only return the first five digits
    elif '-' in code:
        return code[:5]
    # If the postcode does not fall in to these catigories, return None
    else:
        return None
```

For example, the update_postcodes function would make the following changes:

```
02835 => 02835
02903-2996 => 02903
029212 => None, omit record
```

1.1.3 Update Tiger GPS Blocks

While doing a quick through the data, I noticed that there were a lot of ways with their names broken up into segments in a different format. They always had the key start with 'tiger:' which I assumed to mean that they came from an automated upload. Tiger GPS blocks typically have the following format:

```
<tag k="name" v="Commonwealth Avenue" />
<tag k="highway" v="residential" />
<tag k="tiger:cfcc" v="A41" />
<tag k="tiger:county" v="Bristol, RI" />
<tag k="tiger:reviewed" v="no" />
<tag k="tiger:zip_left" v="02806" />
<tag k="tiger:name_base" v="Commonwealth" />
<tag k="tiger:name_type" v="Ave" />
<tag k="tiger:zip_right" v="02806" />
```

You can see that this block has already been collated with a name at the top fo the block in the normal format. During the audit, I decided to check if there were any tiger GPS blocks that had not yet been updated with a name tag. It turns out there were several, and adding the name tag would be useful for standardizing the database. The update_tiger function is used during csv conversion to return the name element. The function also calls the earlier function, update_street_name, to check for abbreviated road types and replace them.

```
def update_tiger(parent_id, tiger_name, tiger_type):
   Function takes in a parent id, tiger name, and tiger type and returns
    a dictionary with the appropriate name dictionary. Should only be called
    when the tiger name has not yet been compiled.
    # initialize a child dict with the appropriate type, key, and id
    child_dict = {'id':parent_id,
                  'type': 'regular',
                  'key':'name'}
    # combine the tiger parts into a full street name
    street = tiger_name + ' ' + tiger_type
    # use the street audit function to fix any abbreviation issues
    street_final = audit_streetnames.update_street_name(street,
                                                    audit_streetnames.mapping)
    # print 'Updated', tiger_name, tiger_type,' => ', street_final
    child_dict['value'] = street_final
   return child dict
```

For example, this function would take in the following block and add a name element.

```
<tag k="tiger:name_base" v="Commonwealth" />
<tag k="tiger:name_type" v="Ave" />
```

This would add the name element:

```
<tag k="name" v="Commonwealth Avenue" />
```

1.2 Database Analysis and Queries

We will start by listing the file sizes for some important files:

```
      rhode-island-latest.osm
      185 MB

      ri_osm.db
      98.9 MB

      nodes.csv
      71.2 MB

      nodes_tags.csv
      2.5 MB

      ways.csv
      5.3 MB

      ways_tags.csv
      12.1 MB

      ways_nodes.csv
      22.8 MB
```

1.2.1 Initial Queries

Next we query to total number of nodes:

```
SELECT count(*) FROM nodes;
```

```
846158
   The total number of ways:
SELECT count(*) FROM ways;
   89540
   The number of distinct users, combining both the nodes and ways database:
SELECT COUNT(DISTINCT(combined.uid)) FROM
(SELECT uid FROM nodes
UNION ALL
SELECT uid FROM ways) combined;
   803
   The total number of shops in the database:
SELECT COUNT(*)
  FROM (SELECT key FROM nodes_tags
 UNION ALL
  SELECT key FROM ways_tags) combined
  WHERE key = "shop";
   412
```

1.2.2 Additional Queries

Now we will compute some additional statistics to explore the database. First, we can look at the most common keys in node_tags to see what is available for exploration.

```
SELECT key, count(*) as num
FROM nodes_tags
GROUP BY key
ORDER BY num DESC;
source
           7119
attribution 6462
          4647
name
          4466
power
ele
          3441
feature_id 3116
           2771
amenity
created
           2564
county id 2373
state_id
           2373
highway
           1947
natural
           1752
```

The most noticable thing is that so many of the keys are for source and other internal documentation use. Of the remaining common keys, amenity and natural seem to be the most interesting to explore.

To start exploring amenities, we will look at the ten most common amenities:

```
SELECT nodes_tags.value, COUNT(*) as num
 FROM nodes_tags
 WHERE key = 'amenity'
  GROUP BY value
  ORDER BY num DESC
 LIMIT 10;
school
                 608
place_of_worship 508
grave_yard
                 410
restaurant
                 177
library
                 116
fire_station
                 112
parking
                  88
fast_food
                  77
bench
                  72
                  63
kindergarten
```

For fun, I decided to query the most common fast food resaurants in RI.

```
SELECT nodes_tags.value, count(*) as num
 FROM nodes_tags
  JOIN (SELECT DISTINCT(id)
        FROM nodes_tags
        WHERE value = 'fast_food') as ff
        ON nodes_tags.id = ff.id
 WHERE nodes_tags.key = 'name'
  GROUP BY nodes_tags.value
  ORDER BY num DESC
 LIMIT 4;
Subway
               13
Dunkin' Donuts 9
McDonald's
Burger King
                3
```

I'm very surprised that Dunkin is not the top. I think most other Rhode Islanders would be too!

Next I looked into the nature tags, and their frequency.

```
SELECT value, count(*) as num
FROM nodes_tags
WHERE key = 'natural'
GROUP BY value
ORDER BY num DESC;

tree 1249
peak 236
bay 128
```

```
beach 73
wetland 51
cliff 10
wood 3
cape 1
rock 1
```

Trees are by far the most common nature tags. I'm surpriesed to see a rock, let's look into what that is!

```
SELECT nodes_tags.value
FROM nodes_tags
JOIN (SELECT DISTINCT(id)
    FROM nodes_tags
    WHERE value = 'rock') as rock
    ON nodes_tags.id = rock.id
WHERE nodes_tags.key = 'name';
```

This query returns **Elbow Rock**.

I'm not totally sure what Elbow Rock is, but a quick search seems to show that there is hiking trail near it.

1.3 Ideas and Issues

This approach for analyzing OpenStreetMap data works well overall, allowing easy creation of the csv files and database. For ideas about where we could improve the database and the data overall, lets take a look at the number of contributions of the top ten users:

```
SELECT user, count(*) AS num
FROM nodes
GROUP BY uid
ORDER BY num DESC
LIMIT 10;
woodpeck_fixbot 338103
greggerm
                212390
                 26848
Zirnch
maxerickson
                 17340
John Wrenn
                 16771
ZeLonewolf
                 16466
morganwahl
                  9693
Roman Guy
                  9432
GeoStudent
                  9419
TIGERcnl
                  8440
```

Looking at the top ten users in nodes, it is clear that some bots are an order of magnitude higher in contribution than some other users. I think it would be very useful to have a table with more information about each user or bot and the kind of data they upload. It might be useful to identify some biases in the data.

The separate user table would have the primary key uid, contain the usernames and other user data such as whether or not it is a bot. For bots, it could also contain some other data provided by the bot writers. Having this separate table would come with several advantages and disadvantages.

The primary benefit of having another table specifically for user information is that it could be used to contain other data. Being able to record data about bot type, status, and the means of data gathering could be useful in assessing the completeness and acuracy of a dataset. For example, perhaps one prolific bot records only a certain type of amenity. This would bias the database toward that type of amenity. The other main advantage of the separate user table is the ability to look at all users in one place, rather than split among the various other nodes and ways tables.

However, splitting the users into another table does come with some drawbacks. First, the table would take up additional memory which we otherwise have free. In addition, having the users in a separate database could increase the complexity of certain queries involving the user names. Storing additional data about users requires care to not reveal any sensitive information and maintain privacy as well. Overall, I think having more user data would benefit the database.

Final Note One other issued that I had occured during csv import to SQL. I got a 'datatype mismatch' error on the first line for the nodes and ways tables only. I'm not sure what in the header causes this error.

The error: nodes.csv:1: INSERT failed: datatype mismatch