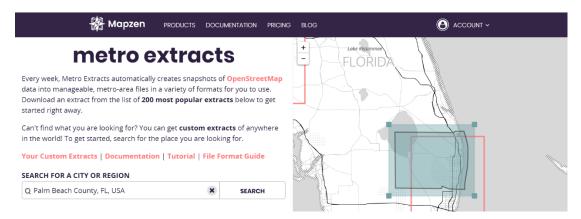
Open Street Data Map Case Study

Map Area:

Palm Beach County, FL

Open Street Map link: https://www.openstreetmap.org/relation/1210748

MapZen Screenshot: The highlighted area in blue is the extract taken for analysis.



I currently live in Palm Beach Gardens and I was interested to explore the data set in the county which I live in i.e. Palm Beach County.

Overview of the Data

Below is a statistics about the data in context:

File Name	File Size
Palm Beach County.osm	192 MB
palmbeach.db	166 MB
nodes.csv	73.5 MB
nodes_tags.csv	1.81 MB
ways.csv	5.62 MB
ways_nodes.csv	25 MB
ways_tags.csv	16.1 MB

Number of unique users:

```
SELECT COUNT(DISTINCT(a.uid))
FROM (SELECT uid FROM nodes UNION ALL SELECT uid FROM ways) a;
```

701

Number of Nodes:

```
SELECT COUNT(*) FROM nodes;
1846632
```

Number of Ways:

```
SELECT COUNT(*) FROM ways;
```

196734

Number of Hospitals in Palm Beach County:

```
SELECT COUNT(*)
FROM (select distinct value from nodes_tags where key='name' and id in (select id from nodes_tags where value='hospital')
UNION
select distinct value from ways_tags where key='name' and id in (select id from ways_tags where value='hospital')) a;
```

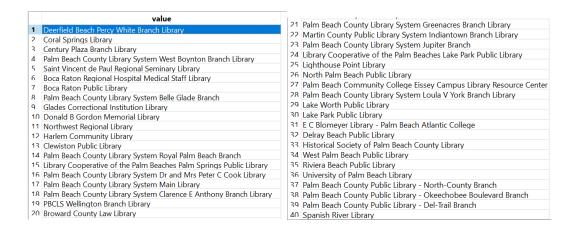
24

Top 10 users who have contributed to Palm Beach County OSM:

```
SELECT b.user, COUNT(*) as num
FROM (SELECT user FROM nodes UNION ALL SELECT user FROM ways) b
GROUP BY b.user
ORDER BY num DESC
LIMIT 10;
woodpeck_fixbot 171056
grouper
             168355
              144088
carciofo
Latze
             119024
Seandebasti
              44492
NE2
              43092
freebeer
              37194
MichaelCollinson 26714
westendguy
              22337
bot-mode
              19135
```

List of Libraries in Palm Beach County:

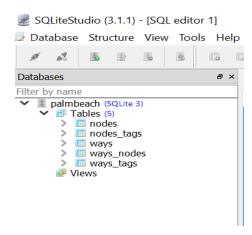
```
!--List of Libraries in Palm Beach County
select distinct value from nodes_tags where key='name' and id in (select id from nodes_tags where key='amenity' and value='library');
```



Problems Encountered in the map extract:

After downloading the osm extract for Palm Beach County, I ran the extract against data.py file to create the following 5 files: nodes.csv, nodes_tags.csv, ways_csv, ways_nodes.csv, ways_tags.csv.

These files were later used to create palmbeach database with the tables as shown in the screenshot below. I used SQLite Studio for this purpose since its open source and very user friendly.



With the help of SQL queries and utilizing the audit.py, following were the main problems encountered:

- *Inconsistent Street Types:* Inconsistences in street types like Street (St, St.), Avenue (Ave), Road (Rd, Rd.), Boulevard (Blvd, Blvd.), Drive (Dr, Dr.)
- *Invalid Postcodes:* Only those postcodes starting with 334 are considered valid Postcodes which fall within Palm Beach County. There are some invalid zip codes which start with FL (Eg: FL 33433) which needs to be updated by removing "FL" from them
- *Inconsistent phone format:* Phone numbers were present in multiple formats. Only phone numbers in xxx-xxx format are considered valid. Some of the other formats were +1 xxx-xxx, (xxx)-xxx-xxxx, xxx xxxx xxxx, etc.
- Data imported from GNIS seem to have old names and they have not been updated. For
 example: "Junior High School" is still widely used for many school names even when it was
 renamed as "Middle School"

Inconsistent Street Types

Utilizing the below code segment from audit.py, all the inconsistent street types were analyzed

```
street_type_re = re.compile(r'\b\S+\.?$', re.IGNORECASE)
expected = ["Street", "Avenue", "Boulevard", "Drive", "Court", "Place", "Square", "Lane", "Road",
      "Trail", "Parkway", "Commons"]
def audit_street_type(street_types, street_name):
   m = street_type_re.search(street_name)
   if m:
       street_type = m.group()
       if street_type not in expected:
          street_types[street_type].add(street_name)
def auditStreet(osmfile):
   osm_file = open(osmfile, "r")
    street_types = defaultdict(set)
    for event, elem in ET.iterparse(osm_file, events=("start",)):
        if elem.tag == "node" or elem.tag == "way":
            for tag in elem.iter("tag"):
# Correct street type
                if is_street_name(tag):
                    audit_street_type(street_types, tag.attrib['v'])
    osm file.close()
    return street_types
```

Based on the above analysis, a mapping Dictionary as shown below was created to fix the inconsistencies.

Utilizing the above mapping dictionary, inconsistent street types are upated

```
def update_name(name, mapping):
    # Updating Inconsistent names
    for key,value in mapping.items():
        print 'Name, Key and Value is ', name," " , key, " ", value
        if key in name:
            updatedName=re.sub(key,value,name)
            print 'Updated Name is: ',updatedName
            break
    return updatedName
```

This updated all inconsistent addresses with street types like St, St., Ave, Rd., Rd, Blvd., Blvd, Dr, Dr. to their corresponding expected Street types.

Invalid Post Codes

Utilizing the below code snippet and SQL query, the formats of postcodes is the osm file were understood. From research it was found that all Palm beach County postcodes start with "334". So, all other postcodes were considered invalid.

```
def audit_post_code(postcode_types, postcode):
    if postcode.startswith("334"):
        postcode_types['valid'].add(postcode)
    else:
        postcode_types['invalid'].add(postcode)

--list of postcodes in Palm Beach County
select distinct value from nodes_tags where key='postcode'
UNION
```

select distinct value from ways_tags where key='postcode';

It was found out that there were multiple instances when postcode was starting with FL (Eg: FL 33433). The below code snippet was used to update these set of incorrect postcodes.

Inconsistent Phone format

Utilizing below code snippet and SQL query, the formats of phone numbers in the osm file were understood. Phone numbers in xxx-xxxx are only considered as in the valid format.

```
def audit_phone(phone_types, phone):
    matchtel=re.match(r'\d{3}-\d{3}-\d{4}',phone)
    if matchtel:
        phone_types['valid'].add(phone)
    else:
        phone_types['invalid'].add(phone)
--list of phone numbers in Palm Beach County
select distinct value from nodes_tags where key='phone'
UNION
select distinct value from ways_tags where key='phone';
```

Some of the major formats generated by the above query included

- XXX-XXX-XXXX
- +1 xxx xxx xxxx
- (xxx) xxx xxxx
- (xxx)-xxx-xxxx

Utilizing the below code snippet some of the inconsistent formats were updated

```
def update_phone(phonetypes):
    #Update phone numbers in the following formats +1 xxx-xxx-xxxx, +1-xxx-xxxx, +1 (xxx)-xxx-xxxx to xxx-xxx-xxxx
    #Update phone numbers in the following format (xxx)-xxx-xxxx, (xxx) xxx-xxxx to xxx-xxx-xxxx
    updatedPhoneList=[]
    for key,value in phonetypes.items():
        #print 'Key and Value is ', key, " ", value
        if key == 'invalid':
            for item in iter(value):
                if item.startswith('+1'):
                    print 'Phone number before update is ', item
                    updatedPhone=re.sub(r'^(\+1\D?)','',item)
                    updatedPhone1=re.sub(r'[\(\\)]','',updatedPhone)
                    updatedPhone2=re.sub(r'\s','-',updatedPhone1)
                    updatedPhoneList.append(updatedPhone2)
                    print 'Updated phone number is ', updatedPhone2
                elif item.startswith('('):
                    print 'Phone number before update is ', item
                    updatedPhone=re.sub(r'[\(\\)]','',item)
                    updatedPhone1=re.sub(r'\s','-',updatedPhone)
                    updatedPhoneList.append(updatedPhone1)
                    print 'Updated phone number is ', updatedPhone1
    return updatedPhoneList
```

Old Names found in GNIS database

In the below osm snippet, it can be seen that the school name is given as "Lake Worth Junior High School". This is an old format and the newer version is "Lake Worth Middle School".

Additional Ideas

Data cleansing automation

Based on the previous analysis of old names found in GNIS database, it was found out that there are lots of old and stale data present in the OSM data set. We can think of a mechanism where bots can be used to validate this data very frequently so that the most valid and recent data is available at any point of time.

Currently, less than 2% of users in Palm Beach county are automated (user bot-mode is the most dominant automated user). This percentage share has to definitely improve so as to retrieve a high quality data from OSM data set.

The possible challenges associated to data cleansing could be deciding on the reliable data source which the bots need to rely on and the frequency of cleansing (since the data set is very huge).

Standardizing keys and values across data set

After exploring the existing data set for various keys and values being used, I found that there is lot of inconsistency on how this data is used across the data set.

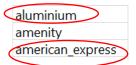
I used the following queries to explore the various keys/values

```
select distinct value from nodes_tags where key='amenity' order by value;
select distinct key from nodes_tags order by key;
```

For example, there are 3 different keys to define cuisine as shown below. This brings lots of inconsistencies in the final data



Another example is:



Key names like "Aluminium" or "American Express" should be avoided and they should be more standardized.

So by standardizing the data, the OSM data set becomes more reliable and more easy to understand for a user who is updating the data.

Conclusion

After exploring the data set, I could find that a major portion of the data for Palm Beach County is fed from TIGER data, produced by the US Census Bureau and GNIS database.

The analysis of the data provided me great insights in to the inconsistencies and invalid data that is present. It also gave me an opportunity to fix some main data prevalent issues which could seriously hamper data analysis. With more standardized data input formats, it would be possible to input a great amount of cleaned data to OpenStreetMap.org.