# Project: Wrangle OpenStreetMap Data - Pierce County WI

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# Introduction

For this project I have chosen to analyze the OpenStreetMap data for Pierce County Wisconsin. I chose this area as my Husband and I plan on moving there in the next 3 years. I will be using data mungling techniques to such as assessing the quality of the data for validity, accuracy, completeness, consistency and uniformity, to clean the OpenStreetMap data. Once the data has been cleand I will use SQL as the data schema for the remainder of the project.

# **Objectives**

Assess the quality of the data for validity, accuracy, completeness, consistency and uniformity.

- Parse and gather data from popular file formats such as .csv, .json, .xml, and .html
- Process data from multiple files or very large files that can be cleaned programmatically.
- Learn how to store, query, and aggregate data using SQL.

# **Auditing the Data**

## **Identify Tag Types**

In order to begin auditing the data I used identifyTags.py to identify the tags used in the datafile. The tags that I will be using for this project are node and way.

A node is a single point in space defined by its longitude, latitude, and node id. A way is an ordered list of nodes.

# Auditing the "k" values

I will begin looking for any problems that may need attention before importing into a database.

First I will look for tags with only lowercase letters, then I will look for lowercase letters separated by a colon, lastly I will look for any problem characters.

```
In [ ]: lower = re.compile(r'^([a-z]|_)*$')
lower_colon = re.compile(r'^([a-z]|_)*:([a-z]|_)*$')
problemchars = re.compile(r'[=\+/&<>;\'"\?%#$@\,\. \t\r\n]')
```

To do this I will use the interparse method of ElementTree to create a dictionary of tags that met those criteria. The full code can be found in auditingK.py.

```
In [ ]: {'lower': 43556, 'lower_colon': 54836, 'other': 2509, 'problemchars': 0}
```

# **Auditing the Users**

For fun I looked into how many users contributed to the map of Pierce County WI using uniqueUsers.py.

I found that 210 unique users made contributions.

# **Problems Encountered**

- The format of the street names is not consistent. Some were abbreviated, some were in uppercase, and some used different abbreviations.
- Zip codes were in inconsistent formats.

## **Street Types**

First issue identified when working with this dataset was the inconsistent use of abbreviations for street types in the street names. In the following code, I create a list of street types that I would expect. Then create a dictionary of the types that are not in my expected list. Using that list I map the abbreviations used in the dataset to the format I specify.

I used the code in streetTypes.py to search through the datafile and compare the values to the expected list. I then print the list of values identified and print them out for review.

```
In [7]:
        import xml.etree.cElementTree as ET
        import pprint
        import re
        from collections import defaultdict
        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 is street name(elem):
            return (elem.attrib['k'] == "addr:street")
        def audit(osmfile):
            osm_file = open(osmfile, "r", errors = 'ignore')
            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"):
                         if is_street_name(tag):
                             audit_street_type(street_types, tag.attrib['v'])
            osm file.close()
            return street_types
        pierce_street_types = audit(datafile)
        pprint.pprint(dict(pierce street types))
        {'47': {'47'},
          '5': {'East Main Street #5'},
          'Ave': {'W Race Ave'},
          'Circle': {'Bauer Circle',
                     'Frederick Circle',
                     'Hackberry Circle',
                     'Melville Circle',
                     'Sandpiper Circle'},
          'E': {'2nd St E'},
          'FLORA': {'CORNER OF WALNUT AND FLORA'},
         'Knoll': {'Highview Knoll'},
          'Ln': {'Learning Ln'},
         'N': {'LAKE ST N'},
         'Path': {'Neill Path'},
         'RD': {'ORRIN RD'},
         'Rd': {'Tyler Rd', 'Twin Bluff Rd'},
          'Rd.': {'Industrial Rd.'},
         'ST': {'BROAD ST'},
          'ST.': {'BROAD ST.', 'DEXTER ST.'},
         'STREET': {'2000 OLD WEST MAIN STREET'},
         'St': {'114th St'},
          'St.': {'N. Maple St.'},
```

'WI-29': {'WI-29'},

```
'Way': {'Village Way', 'Teal Way', 'Sherman Way', 'Glacier Way'}, 'Y': {'County Road Y'}}
```

There are quite a few that are abbreviated in the incorrect format. Now I will use the code in updateStreetTypes.py to update the street names with the name format I identified previously.

## **Zip Codes**

Another problem with this dataset is the zip codes. Some zip codes were in different formats and some were missing completely. First I created a dictionary of the zipcodes that were in an invalid format.

The only issue with the zipcode appears that WI is included in some of them so I corrected those records.

The code for doing hte identification and cleaning of the postal codes can be found in zipCodes.py.

## **Prepare Data for SQL**

Now that the data has been cleaned it's time to prepare the data for loading into SQL.

The XML data will be parsed through and converted into tabular format into CSV files. The CSV files can be imported into sqlite. The code used for this process can be found in sqlPrep.py.

```
In [12]: NODES PATH = "nodes.csv"
         NODE TAGS_PATH = "nodes_tags.csv"
         WAYS PATH = "ways.csv"
         WAY NODES PATH = "ways nodes.csv"
         WAY TAGS PATH = "ways tags.csv"
         LOWER_COLON = re.compile(r'^([a-z]|_)+:([a-z]|_)+')
         PROBLEMCHARS = re.compile(r'[=\+/\&<\;\'''\?\%\#$@\,\. \t\r\n]')
         SCHEMA = schema.schema
         NODE_FIELDS = ['id', 'lat', 'lon', 'user', 'uid', 'version', 'changeset', 'timest
         NODE_TAGS_FIELDS = ['id', 'key', 'value', 'type']
         WAY_FIELDS = ['id', 'user', 'uid', 'version', 'changeset', 'timestamp']
         WAY_TAGS_FIELDS = ['id', 'key', 'value', 'type']
         WAY_NODES_FIELDS = ['id', 'node_id', 'position']
         def shape_element(element, node_attr_fields=NODE_FIELDS, way_attr_fields=WAY_FIELD
                            problem chars=PROBLEMCHARS, default tag type='regular'):
             node_attribs = {}
             way attribs = {}
             way_nodes = []
             tags = [] # Handle secondary tags the same way for both node and way element.
             if element.tag == 'node':
                 for name, value in element.attrib.items():
                     if name in node attr fields:
                          node_attribs[name] = value
                 for secondary in element.iter():
                     if secondary.tag == 'tag':
                          if problem chars.match(secondary.attrib['k']) is not None:
                              continue
                          else:
                              new_dict = tag_dictionary(element, secondary, default_tag_typ
                              if new dict is not None:
                                  tags.append(new dict)
                 return {'node': node attribs, 'node tags': tags}
             elif element.tag == 'way':
                 for name, value in element.attrib.items():
                     if name in way_attr_fields:
                         way attribs[name] = value
                 counter = 0
                 for secondary in element.iter():
                     if secondary.tag == 'tag':
                          if problem chars.match(secondary.attrib['k']) is not None:
                         else:
                              new_dict = tag_dictionary(element, secondary, default_tag_typ
                              if new dict is not None:
                                  tags.append(new dict)
                     elif secondary.tag == 'nd':
```

```
newnd = {}
newnd['id'] = element.attrib['id']
newnd['node_id'] = secondary.attrib['ref']
newnd['position'] = counter
counter += 1
way_nodes.append(newnd)
return {'way': way_attribs, 'way_nodes': way_nodes, 'way_tags': tags}
```

## **Data Overview**

Now that the data is available in SQL I will start exploring it using SQL queries. I will be using the ipython-sql module to connect to the database and run the queries.

#### **Number of Nodes**

## **Number of Ways**

# **Number of Unique Users**

## **Top 5 Contributing Users**

\* sqlite:///om\_pierce\_county.db

### Out[8]: User Name Contributions

jumbanho 117783

woodpeck\_fixbot 46992
iandees 44623
Omnific 39211

PrometheusAvV 17081

#### **Number of Entries**

#### **Number of Cemetaries**

355544

Out[10]: No\_Cemeteries

25

# **Cemetary Names**

```
In [37]:
           %%sql
           SELECT nwt.value as Cemetery_Name
           FROM nodes_tags as nwt
           WHERE nwt.key='name' and nwt.value LIKE '%Cemetery%'
           ORDER BY nwt.value;
            * sqlite:///om_pierce_county.db
Out[37]:
                          Cemetery_Name
                      Beldenville Cemetery
                        Big River Cemetery
                    Diamond Bluff Cemetery
                          Esdaile Cemetery
                 Farm Hill Catholic Cemetery
                      Free Home Cemetery
                 Gilman Lutheran Cemetery
                Greenwood Valley Cemetery
                      Lost Creek Cemetery
                    Maiden Rock Cemetery
                     Maple Grove Cemetery
                Mission Covenant Cemetery
                     Mount Olivet Cemetery
                       Oak Ridge Cemetery
                       Pine Glen Cemetery
                  Plum City Union Cemetery
                       Poplar Hill Cemetery
                       Porcupine Cemetery
            Saint Josephs Catholic Cemetery
                    Saint Josephs Cemetery
                  Salem Lutheran Cemetery
                     Spring Lake Cemetery
                           Tabor Cemetery
                     Thurston Hill Cemetery
```

#### **Number of Post Offices**

Zion Covenant Cemetery

Out[43]: No\_Post\_Office

6

#### **Post Office Names**

\* sqlite:///om pierce county.db

## Out[40]: Post\_Office

East Ellsworth Post Office

Ellsworth Post Office

**Hastings Post Office** 

Red Wing Post Office

River Falls Post Office

Welch Post Office

#### **Number of Cuisines**

```
In [19]:
         %%sql
         SELECT nwt.value as Cuisine, COUNT(*) as Number
         FROM (SELECT * FROM nodes_tags UNION ALL SELECT * FROM ways_tags) as nwt
               INNER JOIN
               (SELECT DISTINCT(id) FROM nodes_tags WHERE nodes_tags.value='restaurant'
               AND nodes_tags.id NOT IN
                                 (SELECT ways_nodes.node_id FROM ways_nodes
                                 (SELECT DISTINCT(id) FROM ways_tags WHERE ways_tags.value=
                                 ON ways_nodes.id = dnwt.id)
               UNION ALL
               SELECT DISTINCT(id) FROM ways_tags WHERE ways_tags.value='restaurant') as d
               ON nwt.id=dwt.id
         WHERE nwt.key='cuisine'
         GROUP BY nwt.value
         ORDER BY Number DESC
         LIMIT 10;
```

Out	[19]	:

Cuisine	Number
american	2
chinese	2
pizza	2
barbecue	1
chicken3	1
coffee_shop	1
ice_cream	1
italian;american;mexican	1
japanese	1

#### **Fast Food Restaurants**

```
In [25]:
         %%sql
         SELECT nwt.value as Restaurant, COUNT(*) as Number
         FROM (SELECT * FROM nodes_tags UNION ALL SELECT * FROM ways_tags) as nwt
                JOIN
               (SELECT DISTINCT(id) FROM nodes_tags WHERE nodes_tags.value='fast_food'
               AND nodes_tags.id NOT IN
                                 (SELECT ways_nodes.node_id FROM ways_nodes
                                 (SELECT DISTINCT(id) FROM ways_tags WHERE ways_tags.value=
                                 ON ways_nodes.id = dnt.id)
               UNION ALL
               SELECT DISTINCT(id) FROM ways_tags WHERE ways_tags.value='fast_food') as wt
               ON nwt.id=wt.id
         WHERE nwt.key='name'
         GROUP BY nwt.value
         ORDER BY Number DESC
         LIMIT 10;
```

#### Out[25]:

Restaurant	Number
Dairy Queen	3
McDonald's	2
Subway	2
Culver's	1
Jimmy John's	1
Randy's	1

# **Additional Ideas**

Since the information is entered in by a variety of users and there is no standardization. To improve this a standard form for entry could be used. That would ensure that each of the amentities, such as restaurants, would have the same information available. This would result in more complete results and better statistics.

	3411cc.///om_pici	
Out[24]:	id	Number
	802228084	2
	802192124	2
	5577469201	5
	5291408314	4
	481461235	3
	458589351	5
	4486234357	2
	4486234327	3
	3821674294	6
	334633471	4
	319416603	3
	319416602	3
	2912759369	7
	2574354284	9
	2491545055	3
	2491545053	5
	2491545051	2
	2491543014	2
	2491543006	2
	2397909513	3
	2397907496	3

2397905736

236315339	4
1833753344	3
125017937	4

## References

OpenStreetMap Wiki: <a href="https://wiki.openstreetmap.org/wiki/Using\_OpenStreetMap">https://wiki.openstreetmap.org/wiki/Using\_OpenStreetMap</a> (<a href="https://wiki.openstreetmap.org/wiki/Using\_OpenStreetMap">https://wiki.openstreetmap.org/wiki/Using\_OpenStreetMap</a>)

Jupyter Magics with SQL: <a href="https://towardsdatascience.com/jupyter-magics-with-sql-921370099589">https://towardsdatascience.com/jupyter-magics-with-sql-921370099589</a>)

GitHub: <a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/tf-coreml/issues/134">https://github.com/tf-coreml/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/issues/134">https://github.com/tf-coreml/issues/134</a> (<a href="https://github.com/tf-coreml/issues/134">https://github.com/tf-c

In [ ]: