OpenStreetMap Data Wrangling Project

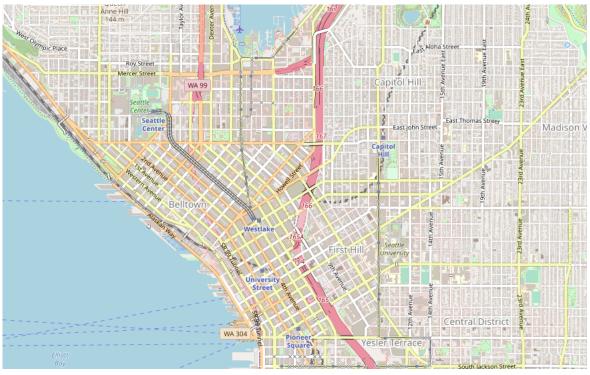
In this notebook I will summarize my steps and findings towards the Data Wrangling Project course, using data obtained from OpenStreet Map.

Introduction

For this project I have chosen to analyze the OpenStreetMap data for Seattle Washington. I chose this area as I currently am living and working in Seattle.

OpenStreetMap Link - https://www.openstreetmap.org/export#map=14/47.6134/-122.3341 (https://www.openstreetmap.org/export#map=14/47.6134/-122.3341)

Out[1]:



Objectives

- · Assess the quality of the data for validity, accuracy, completeness, consistency and uniformity.
- · Parse and gather data
- · Process data
- · Learn how to store, query, and aggregate data using SQL.

I used identifyTags.py to identify the tags used in the datafile.Node and way are the tags I will be looking at.

TagAudit.py was used to look for tags with only lowercase letters, lowercase letters separated by a colon as well as any problem characters.

AuditingK.py was used to find the different attributes represented by the 'k' value and measures their occurance.

UsersUnique.py was used to show how many unique users contributed to the file.

I used StreetTypes.py to audit the street names in the seattle.osm file.

UpdateStreetTypes.py was used to correct inconsistancies with street names in the seattle.osm file.

Tag Identification

This function shows what kind of elements are present in OSM file, and which are important. Reference IdentifyTages.py.

```
In [2]:
            import xml.etree.cElementTree as ET
            import pprint
            OSMFILE = 'Seattle.osm'
            SAMPLE_FILE = 'sample.osm'
            def count tags(filename):
                 tags = \{\}
                 for event, elem in ET.iterparse(filename):
                     if elem.tag not in tags.keys():
                         tags[elem.tag] = 1
                     else:
                         tags[elem.tag] += 1
                 return tags
            tags = count_tags(OSMFILE)
            pprint.pprint(tags)
             {'bounds': 1,
              'member': 70807,
              'meta': 1,
              'nd': 251384,
              'node': 217673,
              'note': 1,
              'osm': 1,
              'relation': 1039,
              'tag': 266030,
              'way': 31137}
```

Size of original file

Sample file creation

The following function creats a sample file. Reference sample.py.

```
In [4]:
         ▶ | k = 25 # Parameter: take every k-th top level element. The value was tuned to
            def get element(filename, tags=('node', 'way', 'relation')):
                context = iter(ET.iterparse(filename, events=('start', 'end')))
                _, root = next(context)
                for event, elem in context:
                    if event == 'end' and elem.tag in tags:
                        yield elem
                        root.clear()
            with open(SAMPLE FILE, 'w', encoding='utf-8') as output:
                output.write('<?xml version="1.0" encoding="UTF-8"?>\n')
                output.write('<osm>\n ')
                # Write every kth top level element
                for i, element in enumerate(get element(OSMFILE)):
                    if i % k == 0:
                        output.write(ET.tostring(element, encoding='utf-8').decode())
                output.write('</osm>')
         | import os
In [5]:
```

```
In [5]: | import os
bytes = os.path.getsize('sample.osm')
mb = float(bytes / 1000000)
print ("osm file size:", mb, "Mb")
osm file size: 2.447847 Mb
```

Auditing the "k" values

The following function finds the different attributes that are represented by the 'k' value and measures their occurance. Reference AuditingK.py.

```
In [6]:
            import pprint
            import xml.etree.cElementTree as ET
            def get types of k attrib(filename, k attrib values dict):
                for _, element in ET.iterparse(filename):
                    if element.tag == "node" or element.tag == "way":
                        for tag in element.iter("tag"):
                            #print(tag.attrib['k'])
                            if tag.attrib['k'] not in k_attrib_values_dict:
                                k attrib values dict[tag.attrib['k']] = 1
                            else:
                                k_attrib_values_dict[tag.attrib['k']] += 1
                            tag.clear()
                        element.clear()
            if __name__ == '__main__':
                k_attrib_values_dict = {}
                filename = "Seattle.osm"
                get_types_of_k_attrib(filename, k_attrib_values_dict)
                #print the top 10 k values appearing in the Seattle.osm file
                import operator
                pprint.pprint(sorted(k attrib values dict.items(),kev = operator.itemgett
            [('addr:housenumber', 22362),
             ('addr:street', 22350),
             ('addr:postcode', 21781),
             ('addr:city', 21772),
             ('highway', 16704),
             ('building', 14789),
             ('name', 9411),
             ('amenity', 5255),
             ('footway', 5177),
             ('crossing', 3959)]
```

The function below will will look for tags with only lowercase letters, then lowercase letters separated by a colon and lastly, any problem characters. Reference TagAudit.py.

```
In [7]:
         from collections import defaultdict
            import pprint
           import re
           import csv
           import codecs
           import schema
            import sqlite3
            import pandas as pd
           lower = re.compile(r'^([a-z]])*$')
            lower_colon = re.compile(r'^([a-z]|_)*:([a-z]|_)*$')
           problemchars = re.compile(r'[=\+/\&;\'''\?%#$@\,\. \t\r\n]')
           def key_type(element, keys):
               if element.tag == "tag":
                       if lower.search(element.attrib['k']) != None:
                           keys['lower'] += 1
                       elif lower colon.search(element.attrib['k']) != None:
                           keys['lower colon'] += 1
                       elif problemchars.search(element.attrib['k']) != None:
                           keys['problemchars'] += 1
                       else:
                           keys['other'] += 1
               return keys
           def process map(file):
               keys = {"lower": 0, "lower_colon": 0, "problemchars": 0, "other": 0}
               for _, element in ET.iterparse(file):
                   keys = key type(element, keys)
               return keys
           process map('Seattle.osm')
   Out[7]: {'lower': 144162, 'lower colon': 116062, 'problemchars': 0, 'other': 5806}
```

The function below finds the number of contributors. Reference UsersUnique.py.

Street Types Audit

The code below audits the street names in the seattle.osm file. Reference StreetTypes.py.

```
expected = ['Street', 'Avenue', 'Boulevard', 'Road', 'Place', 'Parkway', 'Lar
In [9]:
                          'Drive'l
In [10]:
          H
             mapping = { "St": "Street",
                          "St.": "Street",
                          "street": "Street",
                         "Ave": "Avenue",
                         "Ave.": "Avenue",
                          "Blvd": "Boulevard",
                          "Blvd.": "Boulevard",
                          "Boulavard": "Boulevard",
                          "Rd": "Road",
                         "Rd.": "Road",
                          "RD": "Road",
                          "Pl": "Place",
                          "Pl.": "Place",
                          "PKWY": "Parkway",
                          "Pkwy": "Parkway",
                          "Ln": "Lane",
                         "Ln.": "Lane",
                          "Dr": "Drive",
                          "Dr.": "Drive"
```

import pprint

In [11]:

```
import re
from collections import defaultdict
datadir = "data"
datafile = "Seattle.osm"
street_type_re = re.compile(r'\b\S+\.?$', re.IGNORECASE)
expected = ['Street', 'Avenue', 'Boulevard', 'Road', 'Place', 'Parkway', 'Lar
            'Drive']
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
seattle street types = audit(datafile)
pprint.pprint(dict(seattle street types))
{'Alley': {'Post Alley'},
 'Ave': {'1st Ave',
         '2nd Ave',
         '3131 Elliot Ave',
         '34th Ave',
         '3rd Ave',
         '4th Ave',
         '6th Ave',
         'Westlake Ave'},
 'Broadway': {'Broadway'},
 'Court': {'Broadway Court',
           'East Barclay Court',
           'East Conover Court',
           'East Florence Court',
           'East James Court',
           'East Jansen Court',
```

```
'East Remington Court'},
'Driveway': {'West Queen Anne Driveway'},
'East': {'10th Avenue East',
```

While I did see a few very small inconsistancies while auditing the street names, I found the formatting to be mostly clean and consistent. The code below attempts to update the street names. Reference UpdateStreetTypes.py.

```
In [12]:
         import pprint
             import re
             from collections import defaultdict
             datadir = "data"
             datafile = "Seattle.osm"
             street type re = re.compile(r'\b\S+\.?$', re.IGNORECASE)
             expected = ["Street", "Avenue", "Boulevard", "Drive", "Court", "Place", "Squa
                         "Trail", "Parkway", "Commons", "East", "North", "West", "South"]
             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
             seattle_street_types = audit(datafile)
             #Map the abbreviations to the expected types
             MAPPING = { "St": "Street",
                         "ST.": "Street".
                         "STREET": "Street",
                         "ST": "Street",
                         "Rd.": "Road",
                         "Rd": "Road",
                         "RD": "Road",
                         "Ave": "Avenue",
                         "E":"East",
                         "Ln":"Lane",
                         "N": "North"
                         }
             def update_street(name, mapping=MAPPING):
                 m = street type re.search(name)
                 if m.group() in mapping:
```

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```
boundaries = re.compile(r'\b' + m.group() + r'$')
        name = re.sub(boundaries, mapping[m.group()], name)
    return name
for street_type, ways in seattle_street_types.items():
    for name in ways:
        better name = update street(name, mapping=MAPPING)
        print(name, "=>", better name)
West Denny Way => West Denny Way
East Olive Way => East Olive Way
Olive Way => Olive Way
Alaskan Way => Alaskan Way
Newport Way => Newport Way
Martin Luther King Junior Way => Martin Luther King Junior Way
East Denny Way => East Denny Way
East James Way => East James Way
Denny Way => Denny Way
Yesler Way => Yesler Way
East Yesler Way => East Yesler Way
Broadway => Broadway
South Jackson => South Jackson
Westlake Ave => Westlake Avenue
34th Ave => 34th Avenue
2nd Ave => 2nd Avenue
6th Ave => 6th Avenue
3rd Ave => 3rd Avenue
3131 Elliot Ave => 3131 Elliot Avenue
1st Ave => 1st Avenue
4th Ave => 4th Avenue
Post Alley => Post Alley
West Queen Anne Driveway => West Queen Anne Driveway
Vine St => Vine Street
Yesler => Yesler
```

Prepare Data for SQL

The data is now ready to be imported into SQL. The XML data will be parsed through and converted into tabular format into CSV files. The CSV files can then be imported into sqlite. The code used for this process can be found in sqlprep.py.

Audited changes are carried out when converting to CSV in the script below.

```
In [13]:
           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"
              # Create Schema using the schema provided in the project instruction
              SCHEMA = schema = {
                  'node': {
                       'type': 'dict',
                       'schema': {
                           'id': {'required': True, 'type': 'integer', 'coerce': int},
                          'lat': {'required': True, 'type': 'float', 'coerce': float}, 'lon': {'required': True, 'type': 'float', 'coerce': float},
                           'user': {'required': True, 'type': 'string'},
                           'uid': {'required': True, 'type': 'integer', 'coerce': int},
                           'version': {'required': True, 'type': 'string'},
                           'changeset': {'required': True, 'type': 'integer', 'coerce': int}
                           'timestamp': {'required': True, 'type': 'string'}
                      }
                  },
                  'node_tags': {
                      'type': 'list',
                       'schema': {
                           'type': 'dict',
                           'schema': {
                               'id': {'required': True, 'type': 'integer', 'coerce': int},
                               'key': {'required': True, 'type': 'string'},
                               'value': {'required': True, 'type': 'string'},
                               'type': {'required': True, 'type': 'string'}
                           }
                      }
                  },
                  'way': {
                      'type': 'dict',
                       'schema': {
                           'id': {'required': True, 'type': 'integer', 'coerce': int},
                           'user': {'required': True, 'type': 'string'},
                           'uid': {'required': True, 'type': 'integer', 'coerce': int},
                           'version': {'required': True, 'type': 'string'},
                           'changeset': {'required': True, 'type': 'integer', 'coerce': int}
                           'timestamp': {'required': True, 'type': 'string'}
                      }
                  },
                  'way nodes': {
                       'type': 'list',
                      'schema': {
                           'type': 'dict',
                           'schema': {
                               'id': {'required': True, 'type': 'integer', 'coerce': int},
                               'node id': {'required': True, 'type': 'integer', 'coerce': in
                               'position': {'required': True, 'type': 'integer', 'coerce': i
                           }
                      }
                  },
                  'way_tags': {
```

```
'type': 'list',
        'schema': {
            'type': 'dict',
            'schema': {
                'id': {'required': True, 'type': 'integer', 'coerce': int},
                'key': {'required': True, 'type': 'string'},
                'value': {'required': True, 'type': 'string'},
                'type': {'required': True, 'type': 'string'}
            }
        }
    }
}
# Make sure the fields order in the csvs matches the column order in the sql
NODE_FIELDS = ['id', 'lat', 'lon', 'user', 'uid', 'version', 'changeset', 'ti
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']
                Helper Functions
def get_element(osm_file, tags=('node', 'way', 'relation')):
    """Yield element if it is the right type of tag"""
    context = ET.iterparse(osm_file, events=('start', 'end'))
    _, root = next(context)
    for event, elem in context:
        if event == 'end' and elem.tag in tags:
            vield elem
            root.clear()
class UnicodeDictWriter(csv.DictWriter, object):
    """Extend csv.DictWriter to handle Unicode input"""
    def writerow(self, row):
        super(UnicodeDictWriter, self).writerow({
            k : v for k, v in row.items()
        })
    def writerows(self, rows):
        for row in rows:
            self.writerow(row)
def update street(name, mapping):
    street=street_type_re.search(name).group()
    name=name.replace(street, mapping[street])
    return name
#clean element function take tag['value'] and tag['key'] as input and return
def clean_element(tag_value, tag_key):
```

```
## clean street names
    if tag key=='street':
        street type re = re.compile(r'\b\S+\.?$', re.IGNORECASE)
        full addr=tag value
        m = street_type_re.search(full_addr)
        if m:
            street_type = m.group() #group(): Return the string matched by th
            if street_type not in expected:
                if street type in mapping:
                    tag value=update street(full addr, mapping) # call update
    ## return updated tag value
    return tag value
## Clean and shape node or way XML element to Python dict
def shape element(element, node attr fields=NODE FIELDS, way attr fields=WAY)
                  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 ele
    ## clean node element
    if element.tag=='node':
        for primary in element.iter():
            for i in node_attr_fields:
                if i in primary.attrib:
                    node_attribs[i]=primary.attrib[i]
        if len(element)!=0:
            for j in range(0, len(element)):
                childelem=element[j]
                if not problem chars.search(childelem.attrib['k']): ## ignor
                    tag["id"]=element.attrib["id"]
                    tag["type"]=default_tag_type
                    tag['value']=childelem.attrib['v']
                    if ":" in childelem.attrib['k']:
                        k_and_v=childelem.attrib['k'].split(':',1)
                        tag["type"]=k and v[0]
                        tag["key"]=k and v[1]
                        if tag["type"]=='addr':
                            tag["value"]=clean element(tag["value"],tag["key"
                    else:
                        tag["key"]=childelem.attrib['k']
                        if tag["type"]=='addr':
                            print(tag value, tag["key"])
                            tag["value"]=clean_element(tag["value"],tag["key"
                tags.append(tag)
        return ({'node': node_attribs, 'node_tags': tags})
    ## handle way element
    elif element.tag=='way':
        for primary in element.iter():
            for i in way attr fields:
```

```
if i in primary.attrib:
                    way_attribs[i]=primary.attrib[i]
        if len(element)!=0:
            for j in range(0, len(element)):
                childelem=element[j]
                tag={}
                if childelem.tag=='tag':
                    if not problem chars.search(childelem.attrib['k']):
                        tag["id"]=element.attrib["id"]
                        tag["type"]=default tag type
                        tag["value"]=childelem.attrib['v']
                        if ":" in childelem.attrib['k']:
                            k_and_v=childelem.attrib['k'].split(':',1)
                            tag["key"]=k and v[1]
                            tag["type"]=k_and v[0]
                            if tag["type"]=='addr':
                                tag["value"]=clean element(tag["value"],tag["
                        else:
                            tag["key"]=childelem.attrib['k']
                            if tag["type"]=='addr':
                                tag["value"]=clean element(tag["value"],tag["
                    tags.append(tag)
                elif childelem.tag=='nd':
                    #print (childelem.attrib['ref'])
                    way node={}
                    way node['id']=element.attrib['id']
                    way node['node id']=childelem.attrib['ref']
                    way node['position']=j
                    #print(way_node)
                    way nodes.append(way node)
        return ({'way': way attribs, 'way nodes': way nodes, 'way tags': tags
## process the file, clean and write XML into csv according to given schema
def process map(file in):
    """Iteratively process each XML element and write to csv(s)"""
   with codecs.open(NODES_PATH, 'w', encoding='utf-8') as nodes_file, \
         codecs.open(NODE TAGS PATH, 'w', encoding='utf-8') as nodes tags fil
         codecs.open(WAYS_PATH, 'w', encoding='utf-8') as ways_file, \
        codecs.open(WAY_NODES_PATH, 'w', encoding='utf-8') as way_nodes_file,
         codecs.open(WAY TAGS PATH, 'w', encoding='utf-8') as way tags file:
        nodes writer = UnicodeDictWriter(nodes file, NODE FIELDS)
        node_tags_writer = UnicodeDictWriter(nodes_tags_file, NODE_TAGS_FIELD
        ways writer = UnicodeDictWriter(ways file, WAY FIELDS)
        way nodes writer = UnicodeDictWriter(way nodes file, WAY NODES FIELDS
        way_tags_writer = UnicodeDictWriter(way_tags_file, WAY_TAGS_FIELDS)
        nodes_writer.writeheader()
        node tags writer.writeheader()
        ways writer.writeheader()
        way nodes writer.writeheader()
```

```
for element in get_element(file_in, tags=('node', 'way')):
    el = shape_element(element)
    if el:

    if element.tag == 'node':
        nodes_writer.writerow(el['node'])
        node_tags_writer.writerows(el['node_tags'])
    elif element.tag == 'way':
        ways_writer.writerow(el['way'])
        way_nodes_writer.writerows(el['way_nodes'])
        way_tags_writer.writerows(el['way_tags'])

process_map("Seattle.osm")
```

Creating SQL Database

The code below creates the sql database I will be using. Reference sqlCreate.py.

```
In [15]:
          # ref https://stackoverflow.com/questions/50735349/import-csv-into-sqlite3-in
             conn=sqlite3.connect('Seattle.db')
             cur = conn.cursor()
             cur.execute("CREATE TABLE nodes ( id INTEGER PRIMARY KEY NOT NULL, lat REAL,
                 user TEXT, uid INTEGER, version INTEGER, changeset INTEGER, timestamp TEX
             conn.commit()
             node_df = pd.read_csv('nodes.csv', dtype=object)
             node df.to sql('nodes', conn, if exists='append', index=False)
             cur.execute("CREATE TABLE nodes tags ()
                 id INTEGER,\
                 key TEXT,\
                 value TEXT,\
                 type TEXT,\
                 FOREIGN KEY (id) REFERENCES nodes(id)\
             )")
             conn.commit()
             nodetag_df=pd.read_csv('nodes_tags.csv')
             nodetag df.to sql('nodes tags', conn, if exists='append', index=False)
             cur.execute("CREATE TABLE ways ()
                 id INTEGER PRIMARY KEY NOT NULL,\
                 user TEXT,\
                 uid INTEGER,\
                 version TEXT,\
                 changeset INTEGER,\
                 timestamp TEXT\
             )")
             conn.commit()
             way_df=pd.read_csv('ways.csv')
             way_df.to_sql('ways', conn, if_exists='append', index=False)
             cur.execute("CREATE TABLE ways nodes (\
                 id INTEGER NOT NULL,\
                 node id INTEGER NOT NULL, \
                 position INTEGER NOT NULL, \
                 FOREIGN KEY (id) REFERENCES ways(id),\
                 FOREIGN KEY (node id) REFERENCES nodes(id)\
             )")
             conn.commit()
             waynode df=pd.read csv('ways nodes.csv')
             waynode_df.to_sql('ways_nodes', conn, if_exists='append', index=False)
             cur.execute("CREATE TABLE ways tags (\
                 id INTEGER NOT NULL,\
                 key TEXT NOT NULL,\
                 value TEXT NOT NULL,\
                 type TEXT,\
                 FOREIGN KEY (id) REFERENCES ways(id)\
             )")
             conn.commit()
             waytag_df=pd.read_csv('ways_tags.csv')
             waytag_df=waytag_df.dropna(subset=['id', 'key', 'value'], how='any')
             wavtag df.to sql('wavs tags', conn, if exists='append', index=False)
```

Exploring the data further with SQL

Top 5 amenitities

For fun, I took a look at the top 5 amenities in my osm file.

```
■ query = ("SELECT tags.value, COUNT(*) as count \
In [16]:
                          FROM (SELECT * FROM nodes tags \
                         UNION ALL \
                          SELECT * FROM ways_tags) tags \
                         WHERE tags.key='amenity' \
                          GROUP BY tags.value \
                          ORDER BY count DESC \
                          LIMIT 5")
             cur.execute(query)
             top 5 amenities = cur.fetchall()
             print ("Top 5 amenities:\n")
             pprint.pprint(top 5 amenities)
             Top 5 amenities:
             [('bicycle parking', 1400),
              ('parking', 663),
              ('restaurant', 619),
              ('waste basket', 417),
              ('cafe', 362)]
```

This being Seattle, it is not surprising to see cafes in the top 5 amenities.

```
In [17]:
          ▶ | query = ("SELECT tags.value, COUNT(*) as count \
                          FROM (SELECT * FROM nodes tags \
                         UNION ALL \
                         SELECT * FROM ways_tags) tags \
                         JOIN (SELECT DISTINCT(id) FROM (SELECT * FROM nodes tags \
                         UNION ALL \
                         SELECT * FROM ways_tags) WHERE value = 'cafe') as subq \
                         ON tags.id = subq.id \
                         WHERE tags.key = 'street' \
                         GROUP BY tags.value \
                         ORDER BY count DESC \
                         LIMIT 5")
             cur.execute(query)
             top 5 caf street = cur.fetchall()
             print ("Top 5 streets by cafe:\n")
             pprint.pprint(top 5 caf street)
             Top 5 streets by cafe:
```

[('3rd Avenue', 14), ('1st Avenue', 13), ('East Pike Street', 11), ('4th Avenue', 11), ('2nd Avenue', 8)]

It looks like 3rd Avenue in downtown Seattle is the street with the highest number of cafes.

Conclusion

The datafile I used from OpenStreetMap was surprisingly consistant given the size of the file and the number of users. In general, the file was cleaner than expected.

This project was a good way to learn data gathering, auditing, cleaning and analysis. It was also valuable to learn how to create and import data into an SQL database.

I might suggest that OpenStreetMap somehow provide data validation checks to avoid erroneous data entry into the OSM database.

Submission

- A pdf document containing your answers to the rubric questions. This file should document your data wrangling process.
- Your Python code you used in auditing and cleaning your dataset for the final project.
- A text file containing a link to the map position you wrangled in your project, a short description of the area and a reason for your choice.
- An .osm file containing a sample part of the map region you used (around 1 10 MB in size).

Common mistakes and warnings

- If you've been working on this project through a Jupyter Notebook, make sure that your code is transferred to standalone .py scripts before submitting your work. It is recommended that you have one script for each field that you audit, and one script that converts and cleans your data. Make sure that your code is well-commented and organized so that your reviewer can follow your work.
- Your code should carry out cleaning operations on one of the fields that you investigate.
 Make sure that you've performed some programmatic cleaning as part of your project work:
 you don't need to clean all of the problems in your dataset, but you should at least address the most common issues observed.
- Make sure that your scripts carry out the functionality of your cleaning steps. One common
 mistake is to create scripts that target specific audits of the code, but the audited changes are
 not actually carried out when the XML data is converted to JSON or CSV for insertion into
 your chosen database.