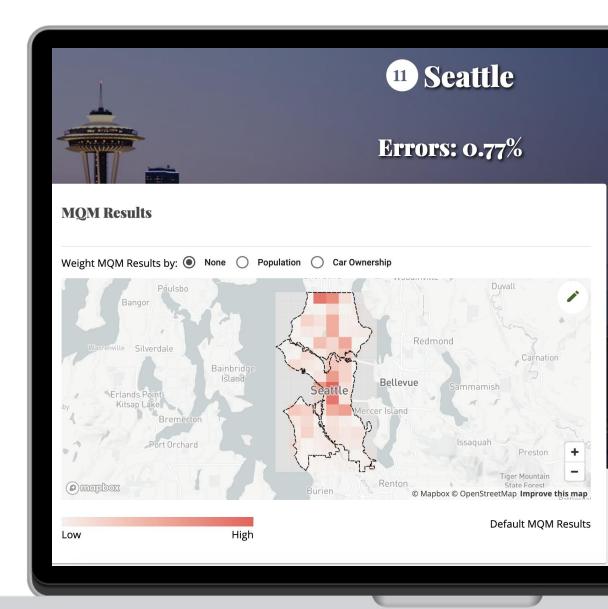
CRITIGEN

MAP QUALITY MEASUREMENTS (MQM)

US Cities Road Data Quality on OpenStreetMap

CUGOS – Fall Fling 2019

Monica Brandeis, Senior GIS/Data Analyst
Critigen



OSM Data Integrity Assurance



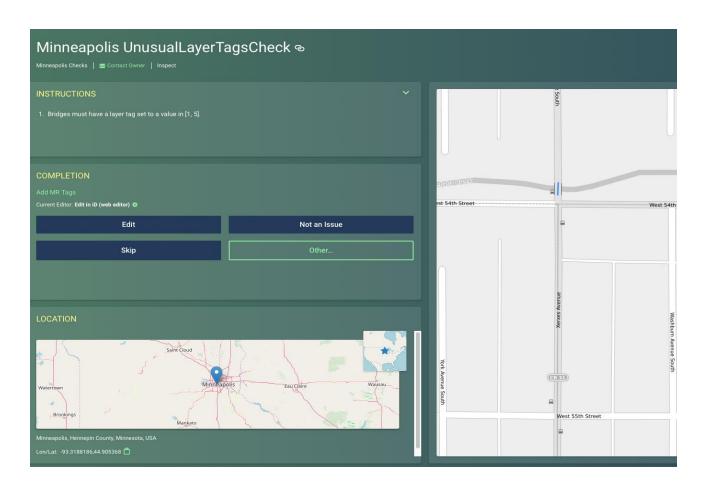
Rule-Based Validation

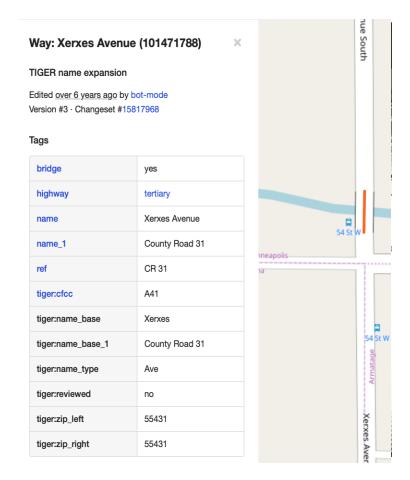
- Finding Data Inconsistency (map errors)
 - Geometry-based
 - Attribute -based

Geometry–Based Map Error



Attribute-Based Map Error





Atlas Checks



A java based program that systematically flags various types of map errors

Atlas Checks



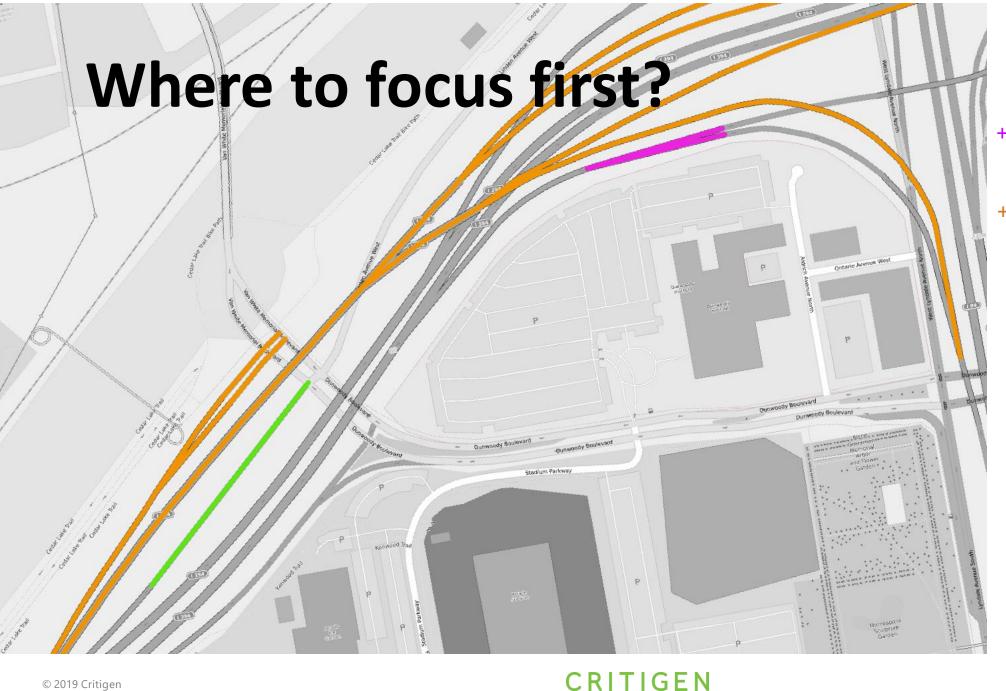
In order to run, need turn OSM data into atlas (a connected graph representation of the road network)



1 OSM Road



5 Atlas Features



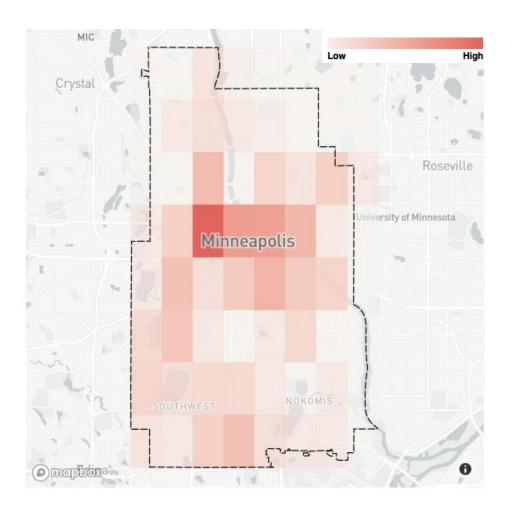
Sign Post Check

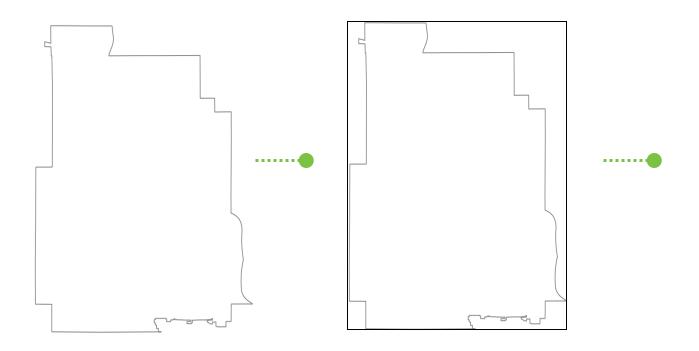
+ Unusual Layer Check

+ Invalid Access Tag Check

Map Quality Measurement (MQM) Tool

- A vector grid layer showing map error hotspots
- Grid size is determined by the distribution of map errors

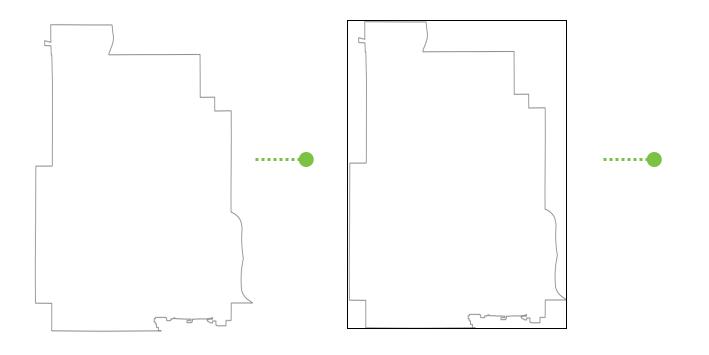




City Boundary

Generate Bounding Box

90% of grids have <10 errors on OSM features 10% of grids (HOTSPOT)



117 117

City Boundary

Generate Bounding Box

Recursively Divide it into half using K-D Tree

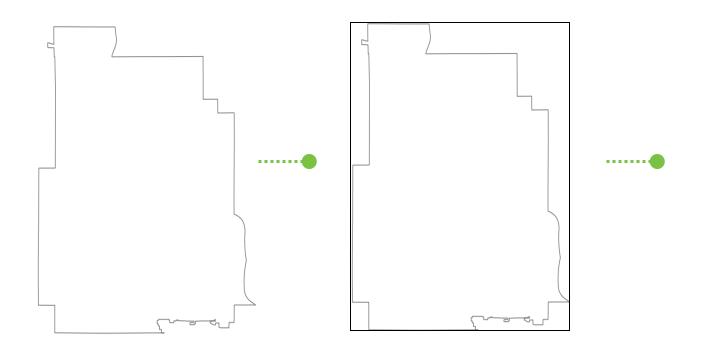


70

City Boundary

Generate Bounding Box

Recursively Divide it into half using K-D Tree

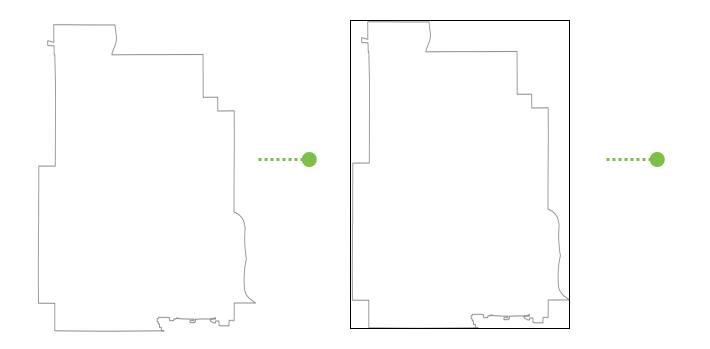


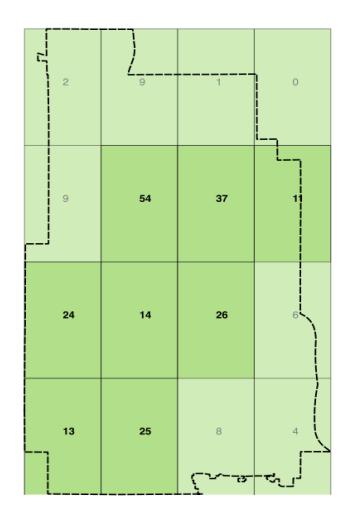
37

City Boundary

Generate Bounding Box

Recursively Divide it into half using K-D Tree

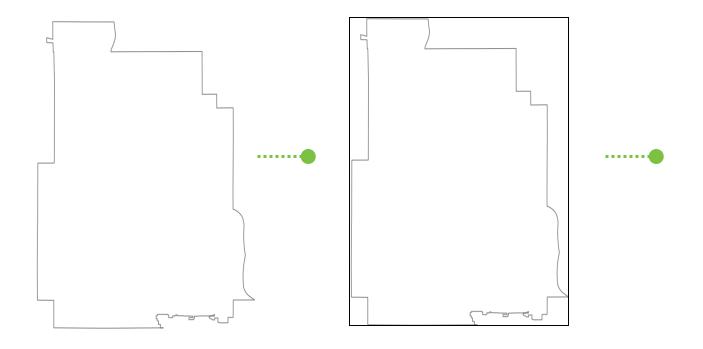


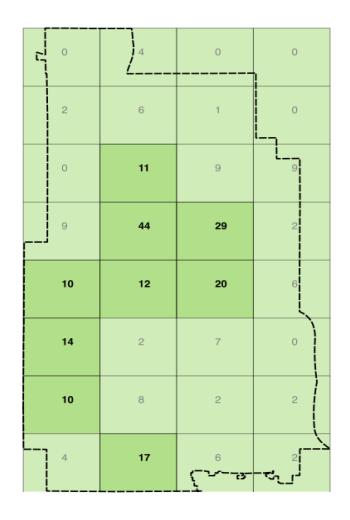


City Boundary

Generate Bounding Box

Recursively Divide it into half using K-D Tree

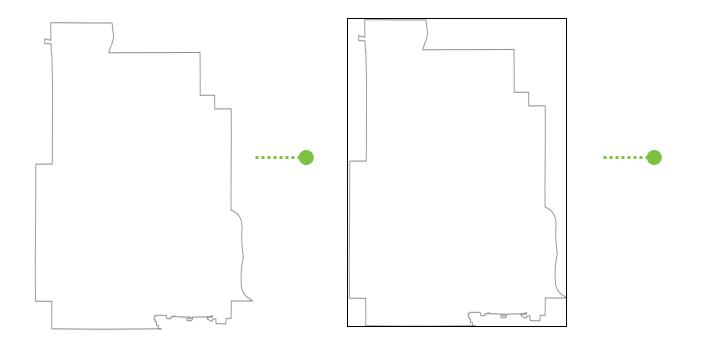


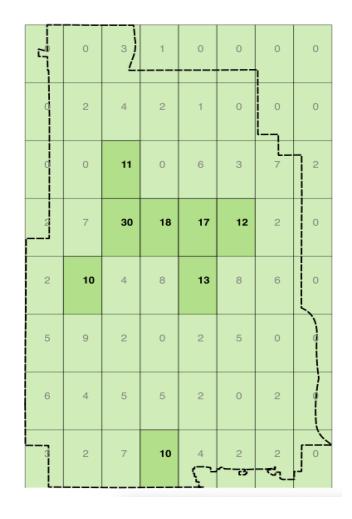


City Boundary

Generate Bounding Box

Recursively Divide it into half using K-D Tree



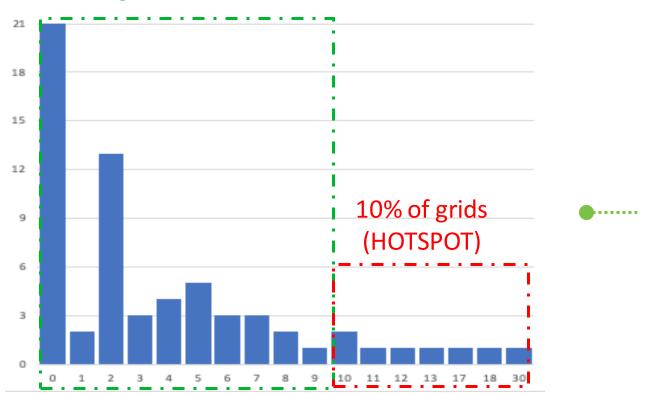


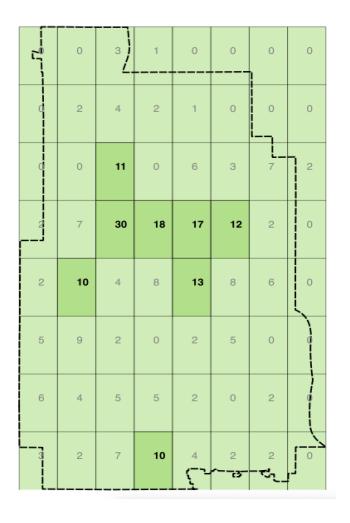
City Boundary

Generate Bounding Box

Recursively Divide it into half using K-D Tree

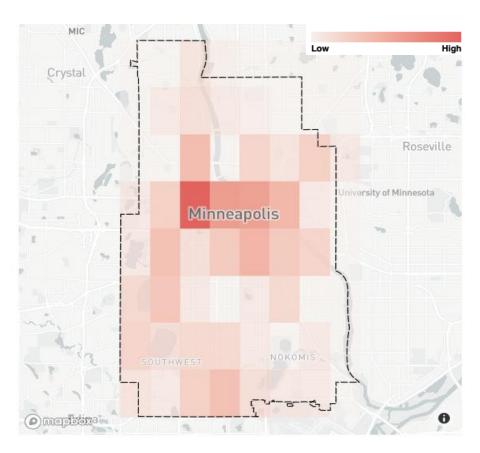
90% of grids have <10 errors on OSM features



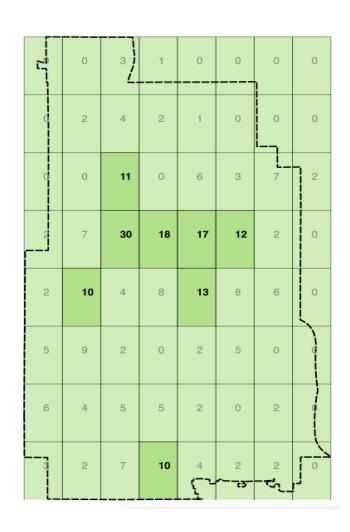


Stop when >=90% of the grids have <10 features

Recursively Divide it into half using K-D Tree



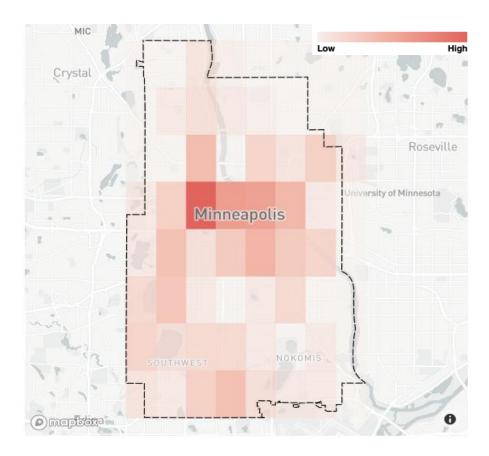
Final MQM Layer



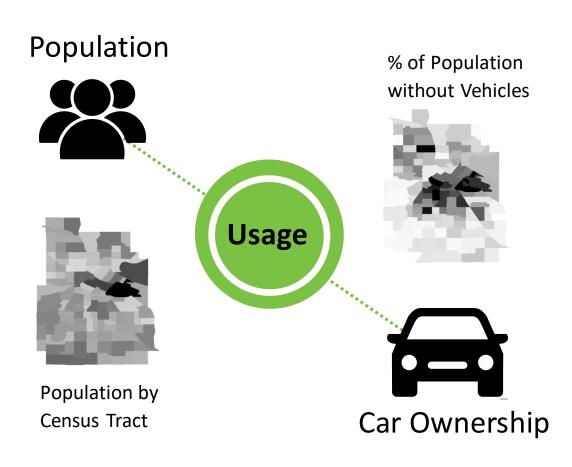
Recursively Divide it into half using K-D Tree

.

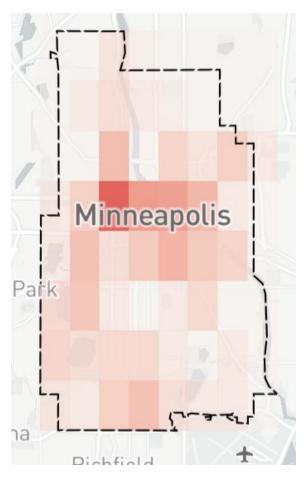
Re-prioritize Map Error Hot-Spots by Usage



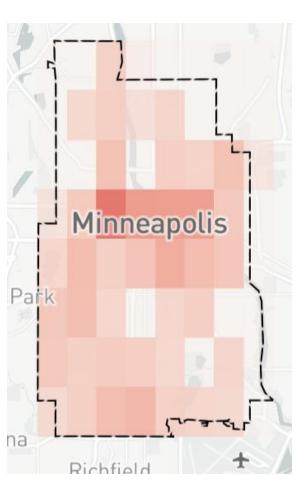
Final MQM Layer



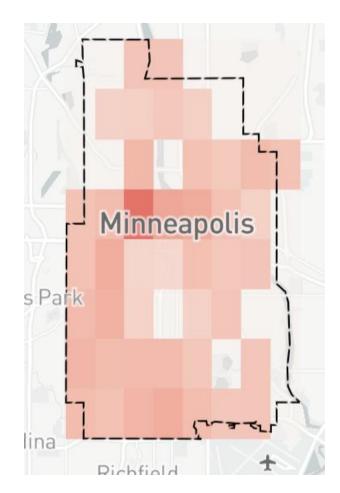
Final 3 Layers



MQM



MQM + Population

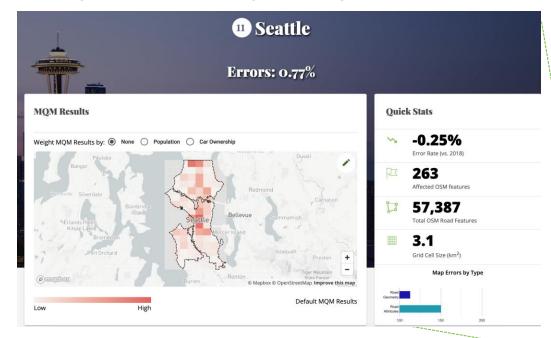


MQM + Car Ownership

CRITIGEN

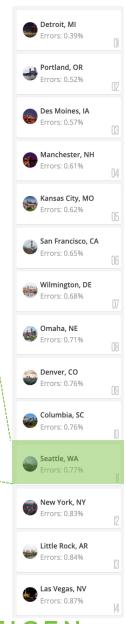
The MQM Web App

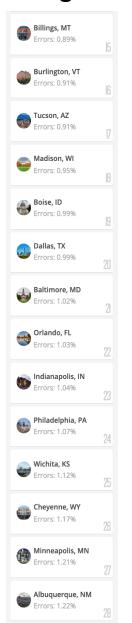
https://osmquality.io

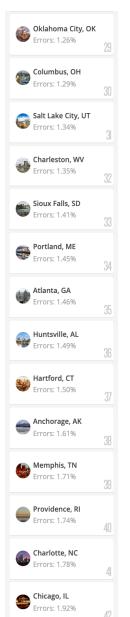


MQM Visualization and Stats for each city

Ranking List for 51 U.S. Cities





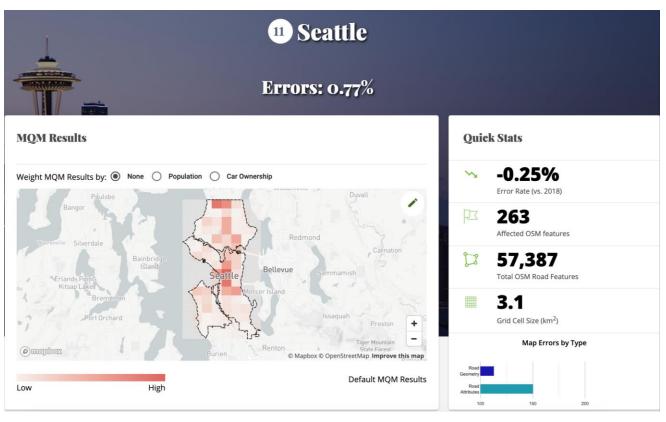




Future Enhancements

- Flexible data inputs (e.g., Overpass Turbo Results)
- Focus on other map layers (e.g., address, building, water, etc.)
- Other measurements for OSM data quality
- Tool incorporation (e.g. JOSM, HOT Task Manager, or MapRoulette)

MQM 2020



More Detail: https://youtu.be/8lpvf9aeyNI

Acknowledgment

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Adam Shaw

Stephen Cerqueira

Daniel Baah

Kim Kearns

Todd Slind





Which city has the best OSM quality

Explore OSM Quality Ranking

Thank you!

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Address
1430 Summit Ave, Suite 100C
Seattle, WA 98122



Selected Atlas Checks for Road Network



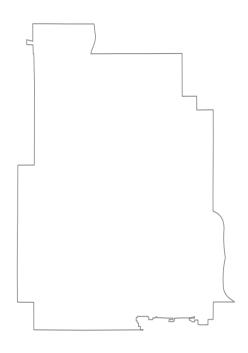
Road Geometry

- EdgeCrossingEdgeCheck
- BuildingRoadIntersectionCheck
- SnakeRoadCheck
- RoundaboutValenceCheck
- InvalidMiniRoundaboutCheck

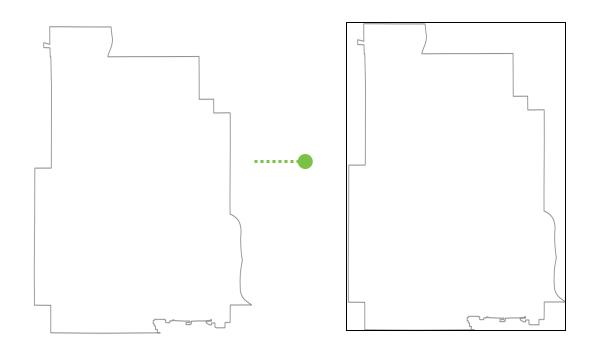


Road Tag and Relation

- SignPostCheck
- InvalidAccessTagCheck
- StreetNameIntegerOnlyCheck
- UnusualLayerTagCheck
- InvalidLanesTagCheck
- InvalidTurnRestrictionCheck



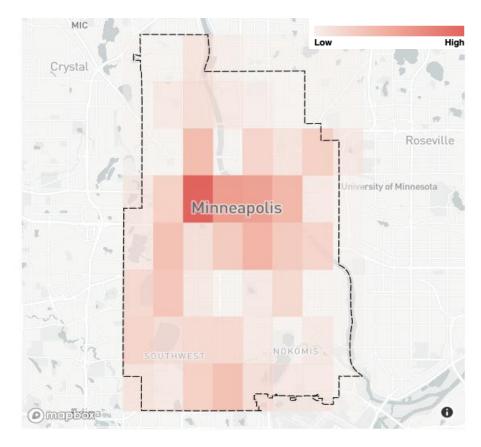
City Boundary



City Boundary

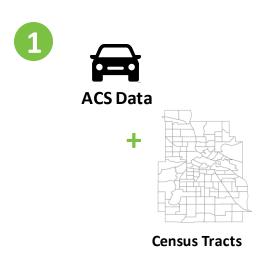
Generate Bounding Box

Re-prioritize Map Error Hot-Spots by Usage



Final MQM Layer

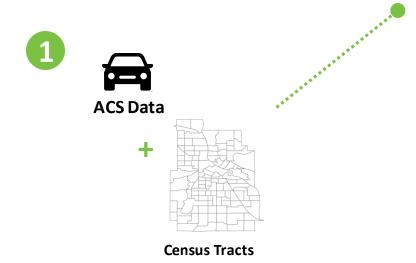
 Join American Community Survey (ACS) data to the census tracts



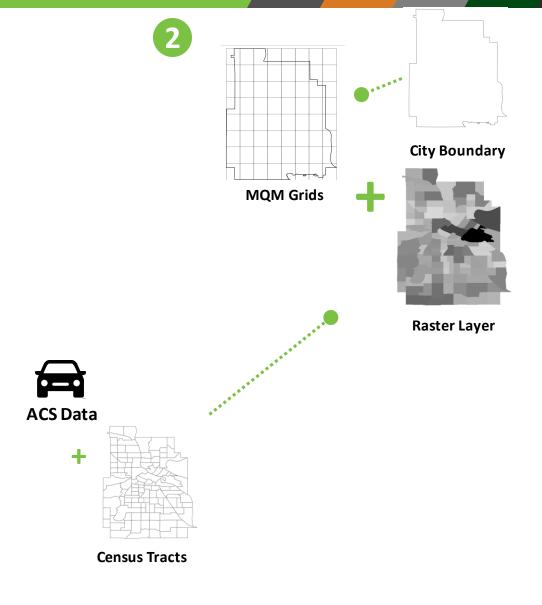
- Join American Community Survey (ACS) data to the census tracts
- Generate raster layers



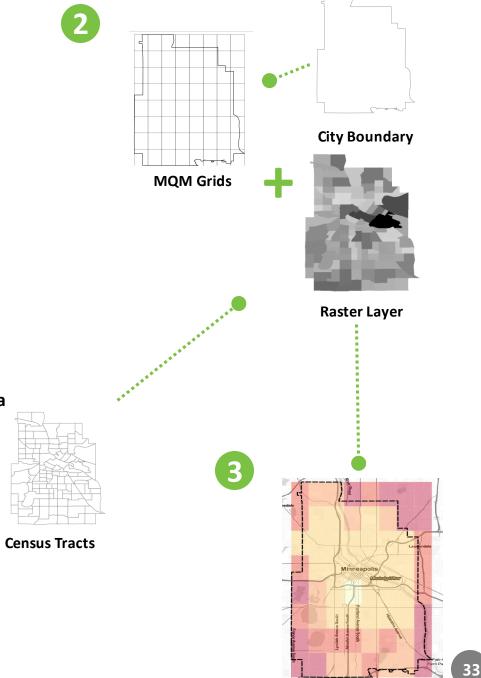
Raster Layer



- Join American Community Survey (ACS) data to the census tracts
- Generate raster layers
- Generate grids using the city boundary

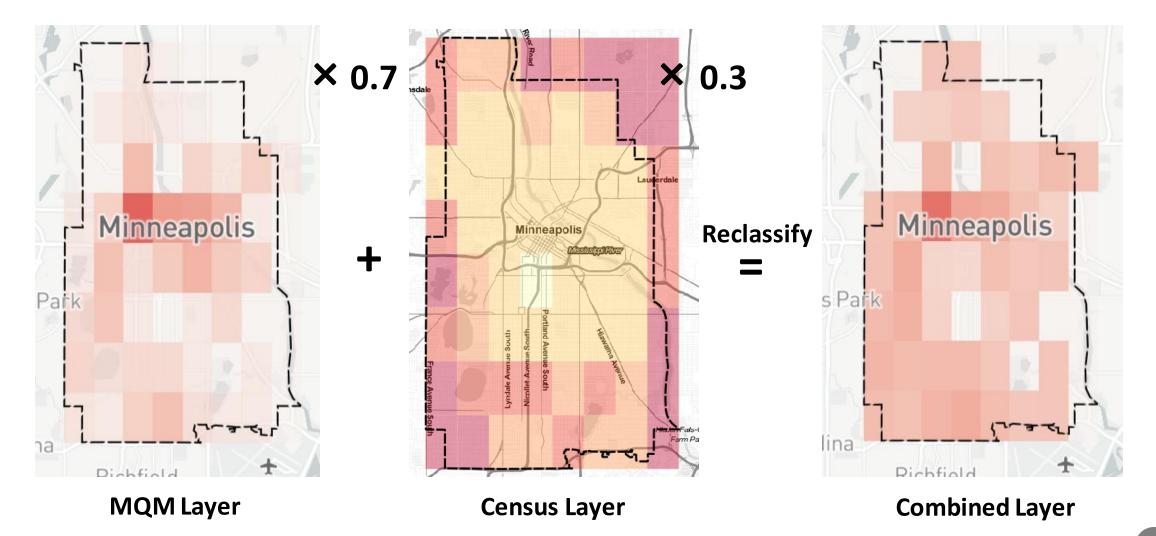


- Join American Community Survey (ACS) data to the census tracts
- Generate raster layers
- Generate grids using the city boundary
- Calculate the mean value of raster layer for each grid
- Normalize values and visualize the results

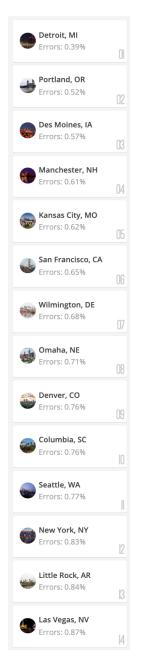


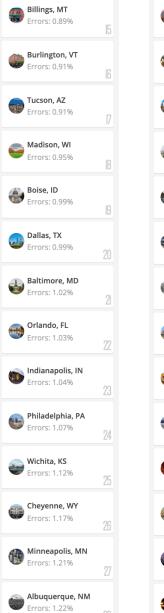
ACS Data

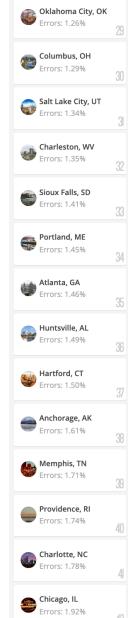
Combining MQM and Census Scores

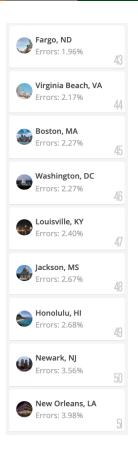


OSM Data Quality Improved Overall



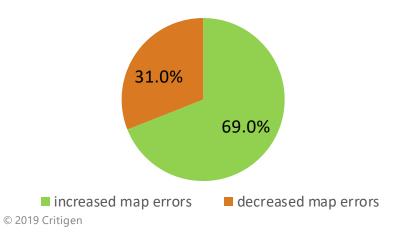






OSM Data Quality Improved Overall

• 69% of the cities (36 out of 51) have a decreased amount of error features





Billings, MT

Burlington, VT

Tucson, AZ

Madison, WI Errors: 0.95%

Boise, ID

Dallas, TX

Baltimore, MD Errors: 1.02%

Orlando, FL

Indianapolis, IN

Philadelphia, PA Errors: 1.07%

Wichita, KS Errors: 1.12%

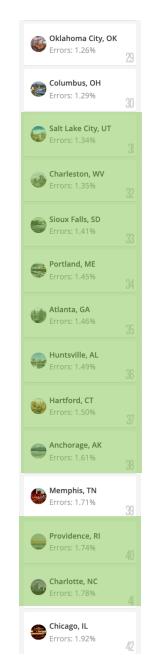
Cheyenne, WY

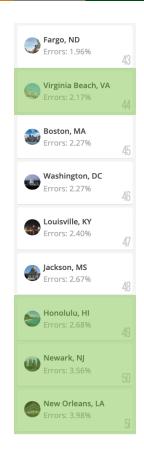
Errors: 1.17%

Minneapolis, MN

Albuquerque, NM

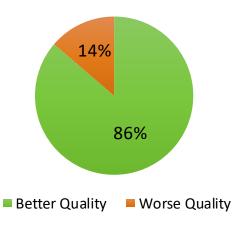
Errors: 1.22%

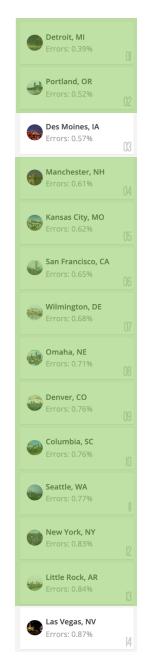




OSM Data Quality Improved Overall

- 69% of the cities (35 out of 51) have a decreased amount of error features
- 86% of the cities (44 out of 51) have a lower MQM Errors rate





Billings, MT

Burlington, VT

Tucson, AZ

Madison, WI Errors: 0.95%

Boise, ID
Errors: 0.99%

Dallas, TX

Baltimore, MD

Orlando, FL

Errors: 1.03%

Indianapolis, IN

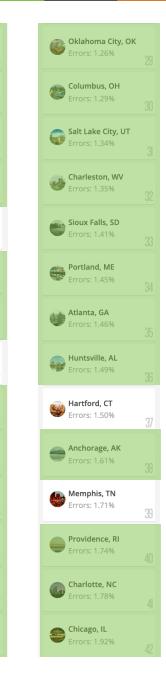
Philadelphia, PA

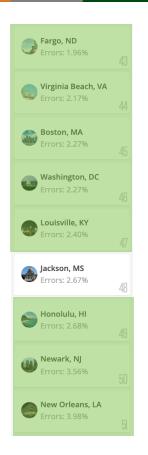
Wichita, KS

Cheyenne, WY

Minneapolis, MN

Albuquerque, NM



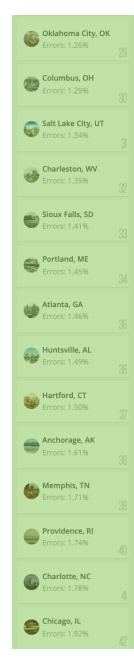


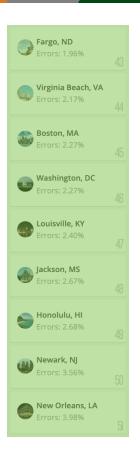
OSM Data Quality Improved Overall

- 69% of the cities (35 out of 51) have a decreased amount of error features
- 86% of the cities (44 out of 51) have a lower MQM Errors rate
- The total feature counts increased in each city

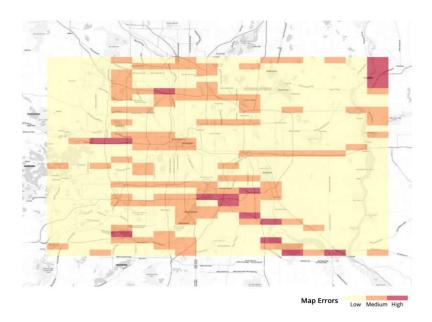


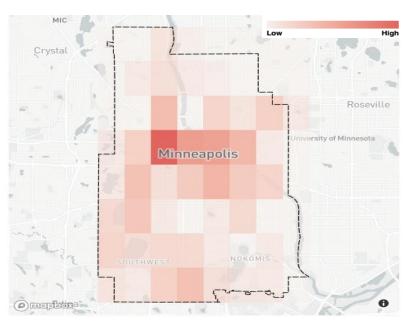




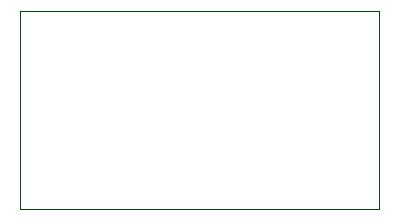


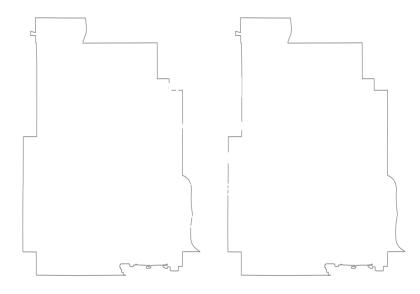
More accurate representations of cities



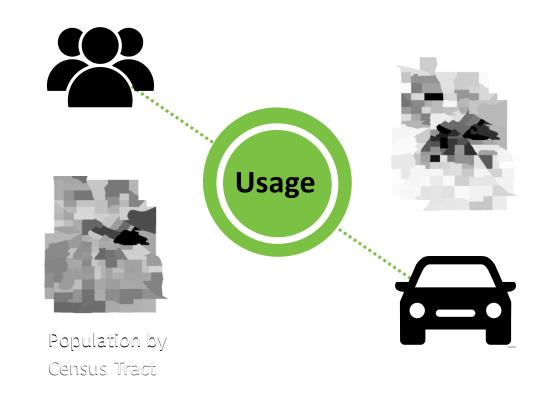


- More accurate representations of cities
- Different OSM data source





- More accurate representations of cities
- Different OSM data source
- Map error hot-spots reprioritization by usage

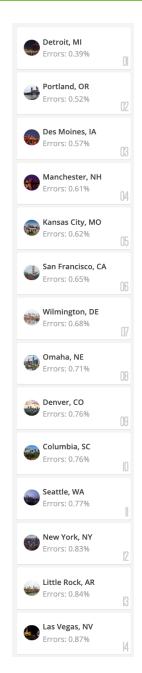


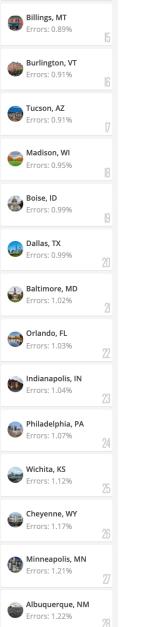
- More accurate representations of cities
- Different OSM data source
- Map error hot-spots reprioritization by usage
- Trend Analysis between 2018 and 2019

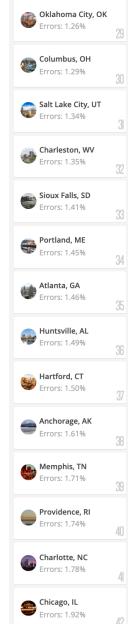


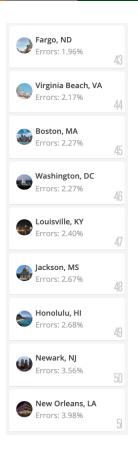
City Rankings

- Rankings are based on errors rates
- Error rate = $\frac{\#error\ features}{\#total\ features} \times 100\%$
- The error rate is an estimation of the percentage of road features that have mapping errors
- A city with increasing error features might rank higher than a city with decreasing error features









OSM Data Quality Improved Overall

- 69% of the cities (35 out of 51) have a decreased amount of error features
- 86% of the cities (44 out of 51) have a lower MQM Errors rate
- The total feature counts increased in each city
- The MQM Errors rate at each ranking decreased

