



Leibniz Institute of
Ecological Urban and
Regional Development



**STATE OF
THE MAP**
FIRENZE

State of the Map 2022
Florence, Italy

Automated derivation of public urban green spaces via activity-related barriers using OpenStreetMap.

Theodor Rieche



About

Theodor Rieche



- cartographer & spatial scientist
- research associate at IOER, research area “Spatial Information and Modelling” (since December 2021)
- currently working in project GOAT 3.0 (Geo Open Accessibility Tool)
- interests: Open Source&Open Data, OSM, Spatial AI, Citizen Science ...

Master thesis

- supervisors

- Prof. Dr.-Ing. **Martina Müller** (University of Applied Sciences Dresden / HTW)
- Dr.-Ing. **Robert Hecht** (Leibniz Institute of Ecological Urban and Regional Development Dresden / IOER)



Project „MeinGruen – Information and navigation to urban green spaces“

- Research project (2018-2021)
- Funded by BMVI (mFUND)
- Public urban green spaces
- „meinGrün“-App
- Filter by criteria or activity possible
- Pilot cities Dresden and Heidelberg
- Polygon base to store features

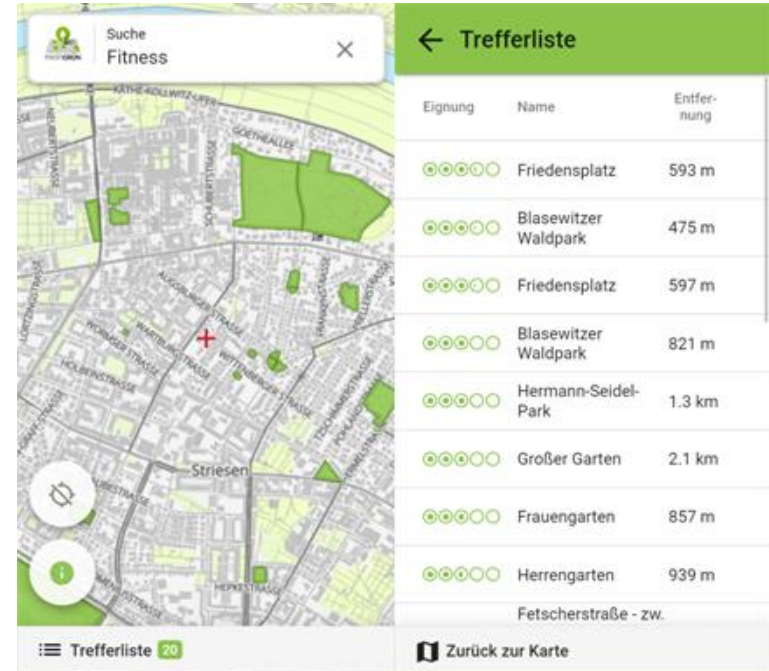


Fig. 1: Screenshot „meinGrün“-App [1]

[1] <https://meingruen.ioer.info/>

Motivation

- Incomplete data of urban green spaces in official data
- Green spaces are missing (are more than parks)
- Different data sources → different data licenses → only OSM possible?
- Consideration of the reality of life / perception of the users of green spaces?
- Test of models to predict greenness or publicly accessibility?

Research questions

- How well is OpenStreetMap data suited for deriving publicly accessible green spaces in urban areas?
- Which land use transitions or key-value (object type) mapped in OpenStreetMap have which probability of being a barrier?

Study area

Pilot city Dresden (Germany) + 5 km buffer

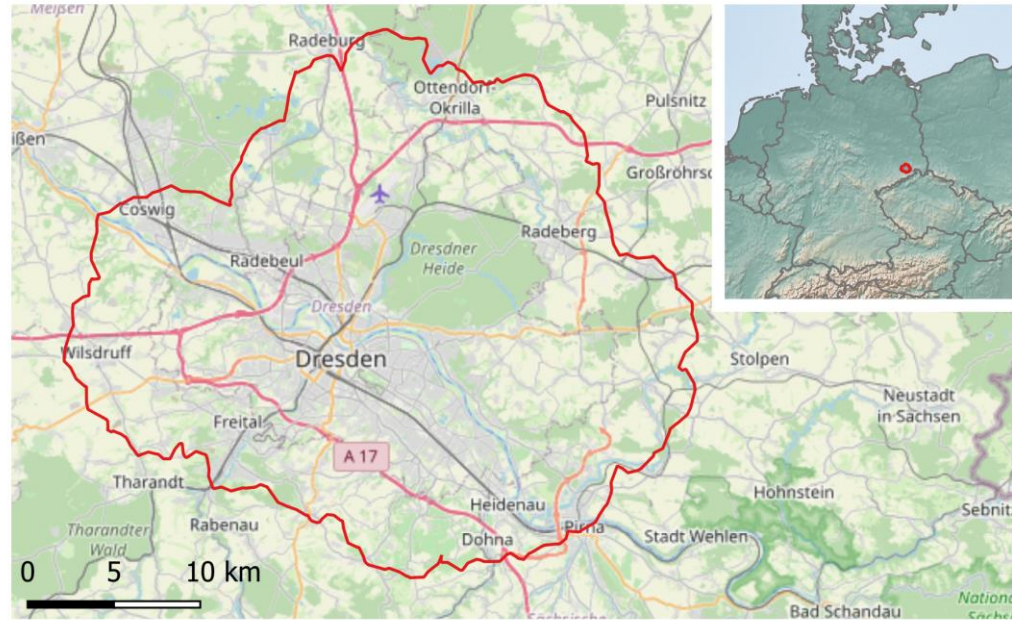


Fig. 2: Study area Dresden, Germany [2]

Data sources

- OpenStreetMap
 - Streets, railroads, waterways, barriers, land use, ...
- official cadastral data (ALKIS) → having field "TN" / land use
- cadastral parcels owned by the city of Dresden
 - Requested in the city council of the city of Dresden



Definition of an activity-related barrier

- Physical barriers such as walls, fences, hedges (barrier=*)
- Action space of doing an activity → delimited by barriers
- Activities divided into „stationary“ or „in motion“
- Roads, railroads, waterways → are always barriers
- Trails or change of land use → uncertain knowledge of being a barrier

Ergo

- To simplify the model → reduction to stationary activities
- Conceptual framework extends OSM definition of barrier

Definition of a activity-related barrier

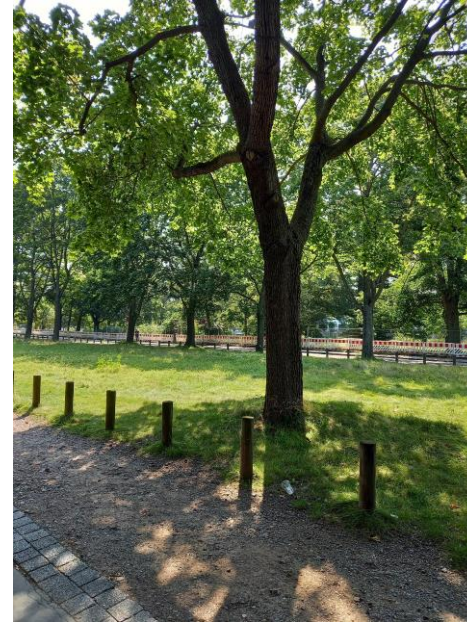


Fig. 3/4: Examples for barriers (flowerbeds, bollards)

Methodology

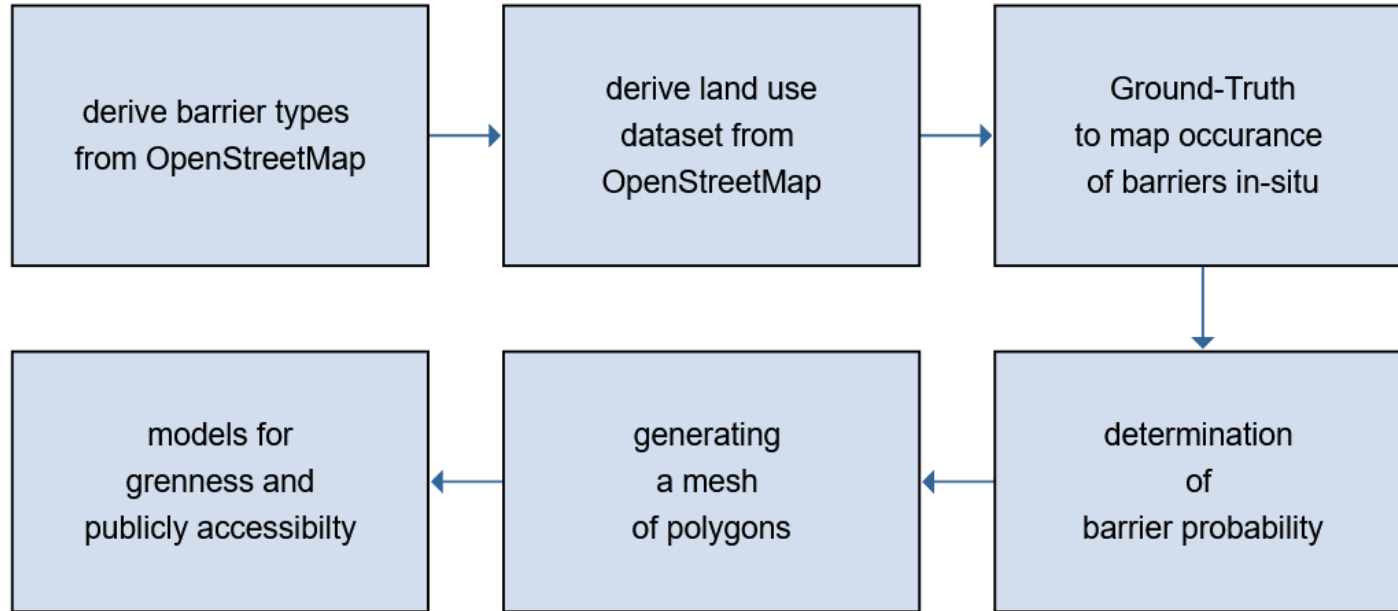


Fig. 5: conceptual framework

Technical implementation

- Ubuntu 20.04 LTS
- dev environment based on Docker container
- Each container having specific installed packages
- PostgreSQL/ PostGIS-database
- SQL, PL/pgSQL, Python, Jupyter Notebook
- Open Source approach

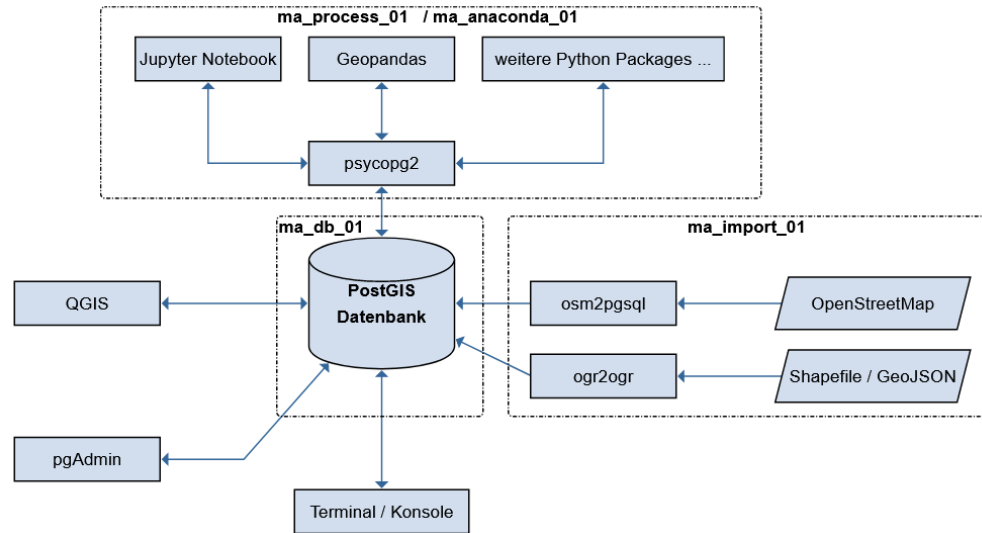


Fig. 6: Technical implementation

Derivation of barrier types from OpenStreetMap

- Roads
- Railroads
- Waterways
- Barriers
- Trails
- Change of land use

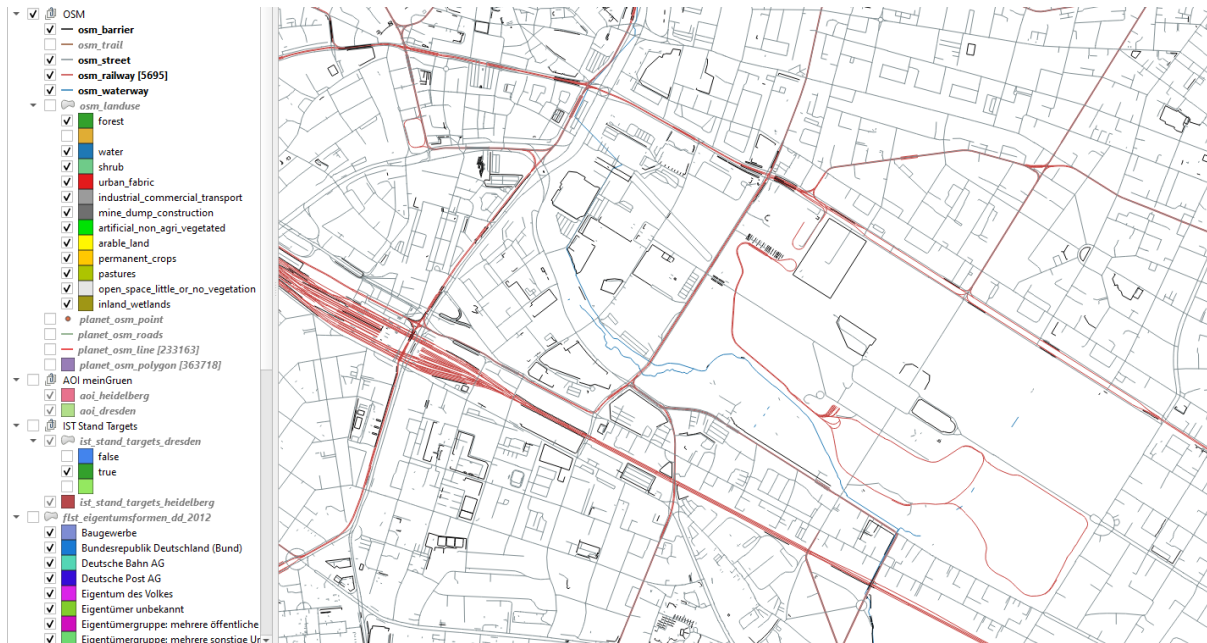


Fig. 7: derived barriers (Dresden, Germany)

Certain and uncertain knowledge

| osm_railway | osm_trail | osm_street | osm_waterway | osm_barrier |
|---|---|--|---|--|
| railway='construction' railway='disused' railway='facility' railway='funicular' railway='miniature' railway='narrow_gauge' railway='platform' railway='platform_edge' railway='preserved' railway='rail' railway='tram' railway='tram_stop' railway='turntable' | highway = 'bridleway' highway = 'cycleway' highway = 'footway' highway = 'no' highway = 'path' highway = 'track' | highway = 'construction' highway = 'living_street' highway = 'motorway' highway = 'motorway_link' highway = 'pedestrian' highway = 'platform' highway = 'primary' highway = 'primary_link' highway = 'raceway' highway = 'residential' highway = 'road' highway = 'secondary' highway = 'secondary_link' highway = 'service' highway = 'steps' highway = 'tertiary' highway = 'tertiary_link' highway = 'trunk' highway = 'trunk_link' highway = 'unclassified' | waterway = 'canal' waterway = 'dam' waterway = 'ditch' waterway = 'drain' waterway = 'fish_pass' waterway = 'river' waterway = 'stream' | barrier=* (all values are relevant) |
| Remove bridges and tunnels. highway=elevator, only if no closed line (to avoid indoor elevators) | Remove bridges and tunnels. highway=steps, check adjacent highway-key | Remove bridges and tunnels. highway=steps, check adjacent highway-key | Remove bridges and tunnels. also tunnel=culvert | Applied to osm „polygons“ and „lines“. Also convert „poylgons“ to „lines“. |

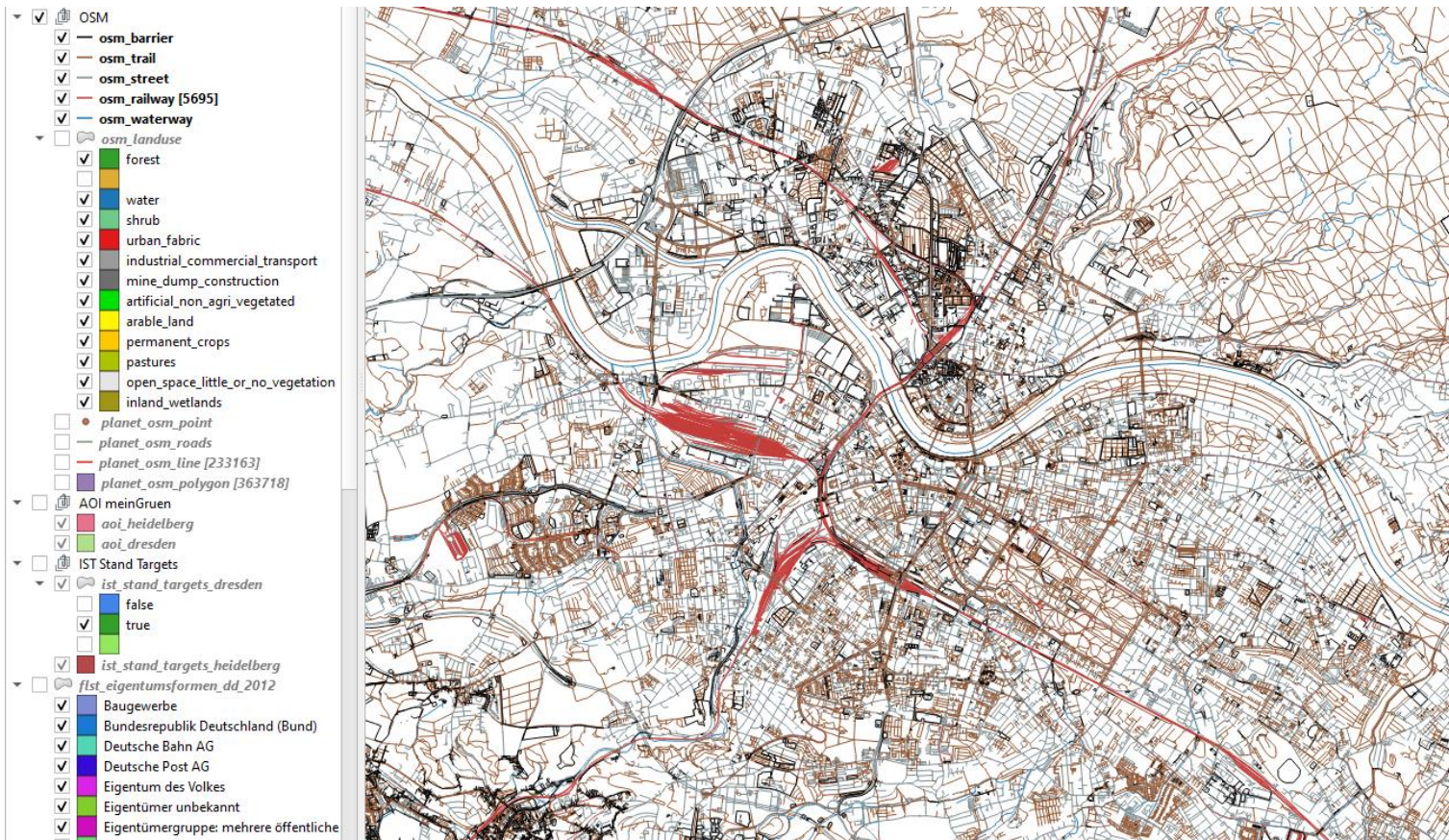


Fig. 9: Screenshot QGIS showing derived barriers

Derivation of land use layer (without overlaps and holes)

- To extract land use changes as lines; also using a residual class

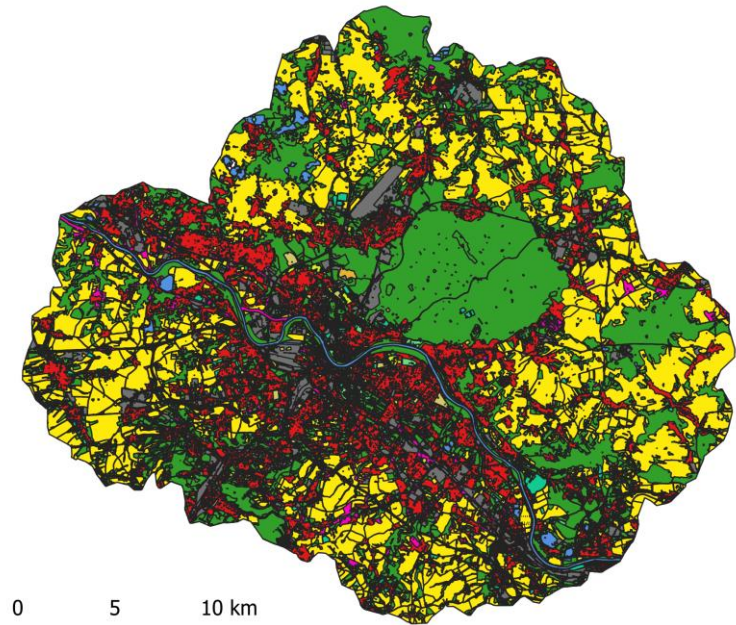
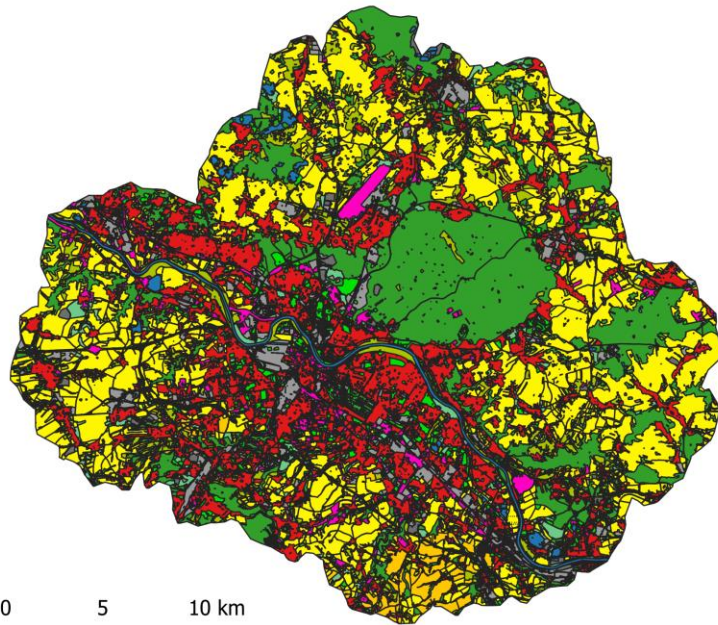


Fig. 10/11: option A (GIScience / osmlanduse.org) and option B (IOER Monitor)

- OSM
 - osm_barrier
 - osm_trail
 - osm_street
 - osm_railway [5695]
 - osm_waterway
 - osm_landuse
 - forest
 - water
 - shrub
 - urban_fabric
 - industrial_commercial_transport
 - mine_dump_construction
 - artificial_non_agri_vegetated
 - arable_land
 - permanent_crops
 - pastures
 - open_space_little_or_no_vegetation
 - inland_wetlands
 - planet_osm_point
 - planet_osm_roads
 - planet_osm_line [233163]
 - planet_osm_polygon [363718]
 - AOI meinGruen
 - aoi_heidelberg
 - aoi_dresden
 - IST Stand Targets
 - ist_stand_targets_dresden
 - false
 - true
 -
 - ist_stand_targets_heidelberg
 - f1st_eigentumsformen_dd_2012
 - Baugewerbe
 - Bundesrepublik Deutschland (Bund)
 - Deutsche Bahn AG
 - Deutsche Post AG
 - Eigentum des Volkes
 - Eigentümer unbekannt
 - Eigentümergruppe: mehrere öffentliche

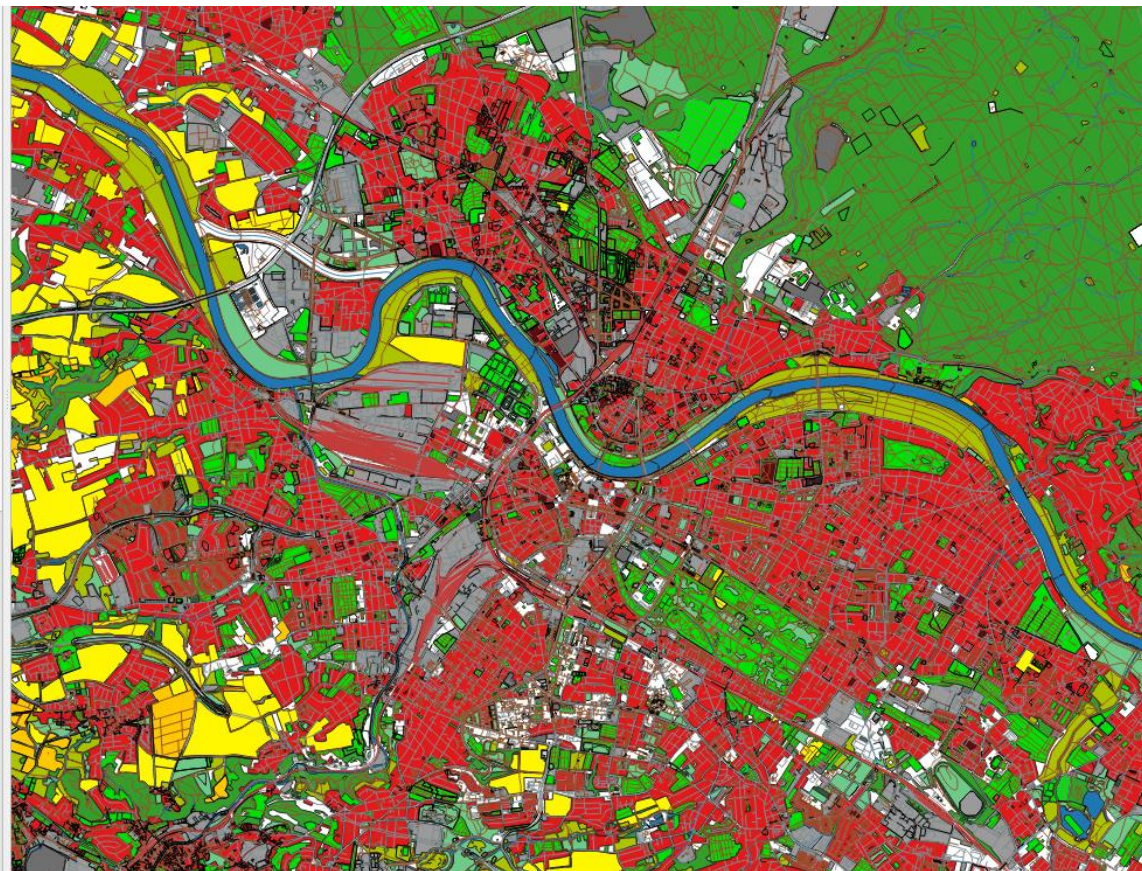


Fig. 12: Screenshot QGIS showing land use

Ground-Truth in-situ mapping

Only for trails and land use changes

- Goal: generate knowledge about being a barrier by type
- QField-App with prepared forms, Barrier: „yes“, „no“, „nodata“

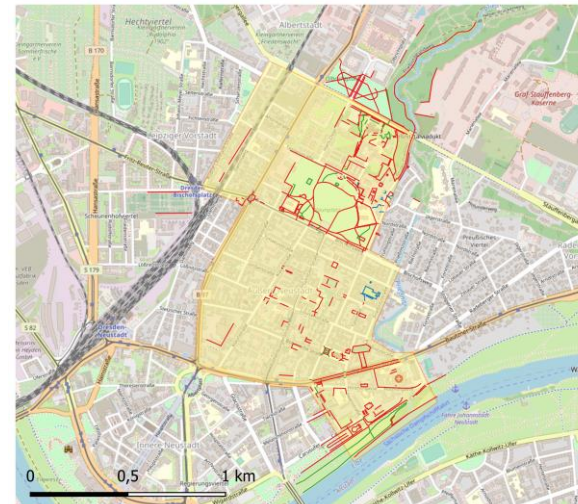


Fig. 13/14: mapped barriers (city park area and new town area in Dresden)

Ground-Truth in-situ mapping

Mapped objects:

| area | trails | | land use changes | |
|----------------------|--------|-----------------|------------------|-----------------|
| | number | Σ length | number | Σ length |
| city park | 297 | 64.682,28 m | 1.145 | 41.429,90 m |
| new town | 96 | 12.802,35 m | 548 | 19.718,66 m |
| miscellaneous | 15 | 4.816,81 m | 27 | 3.057,84 m |
| sum | 408 | 82.301,44 m | 1.720 | 64.206,40 m |

Fig. 15: mapped line segments

- Barrier probabilities were calculated for each type of trail or land use change
- Weighted by length

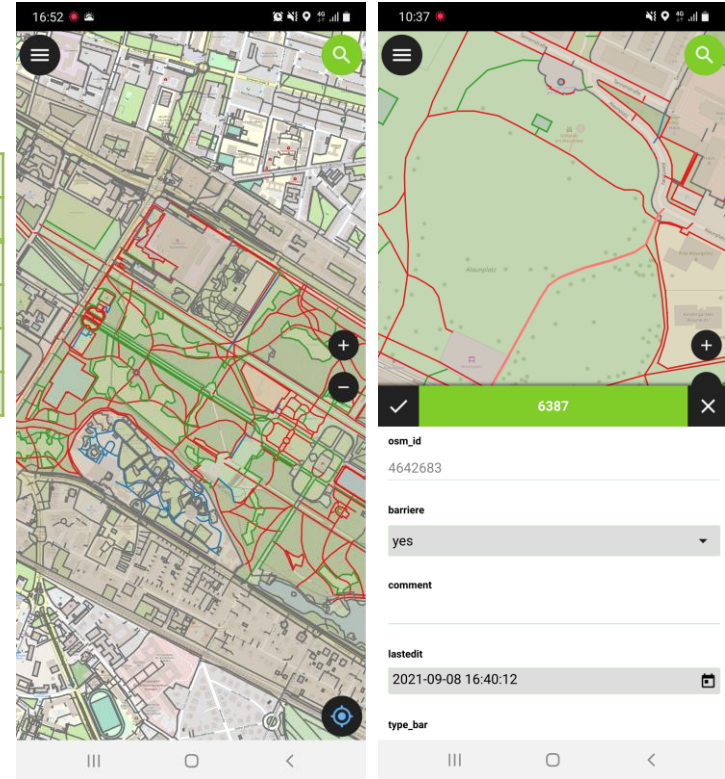
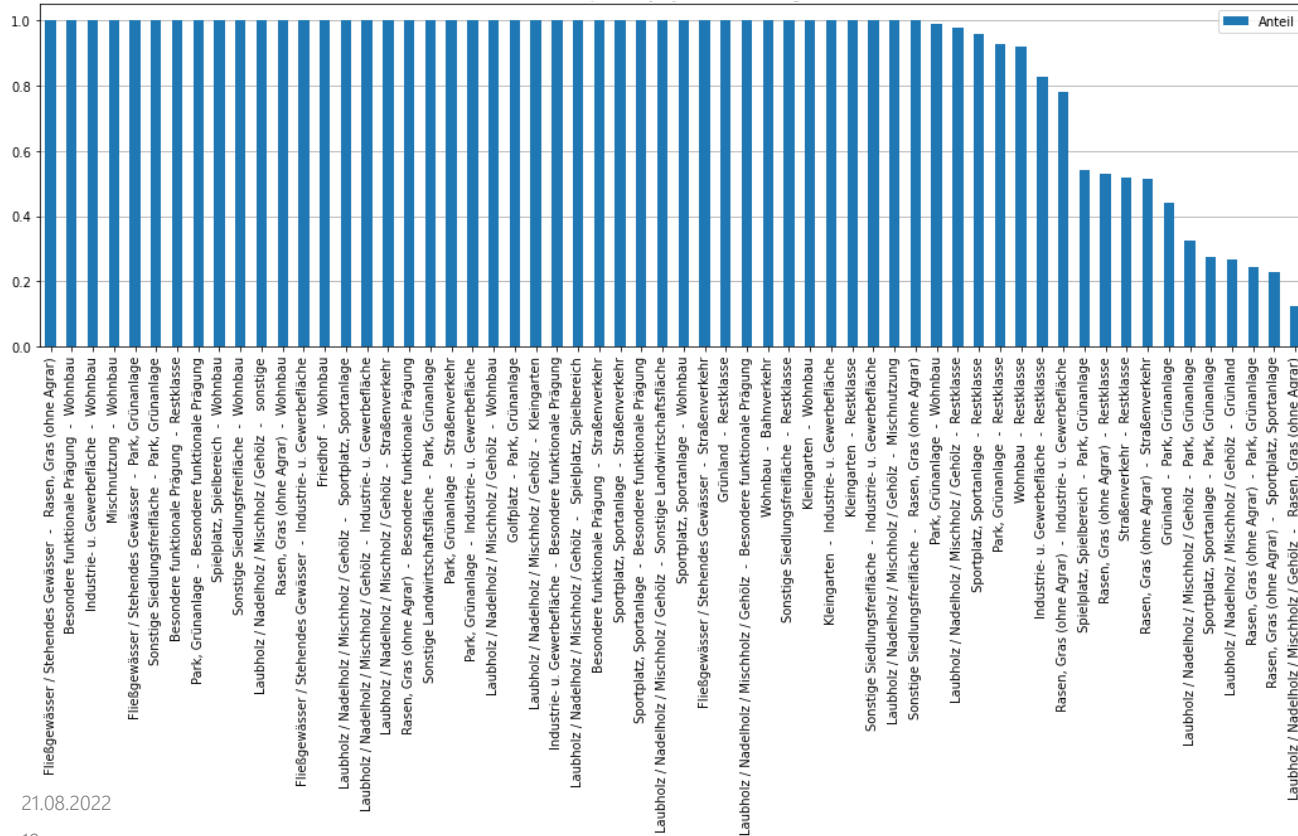


Fig. 16: screenshots QField app

Results of barrier probabilities

barrier probability by land use change



barrier probability by trail type

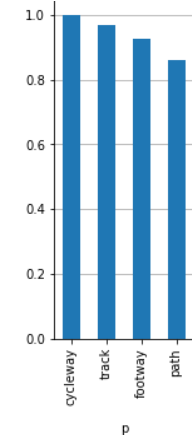


Fig. 17/18: barrier prob. for land use change and trail type

(Unfortunately only with german labels)

Generating a polygon mesh

- First: creating a dataset of all lines („line pool“)

| Barrier type | Additional attributes for each line | | |
|------------------------|-------------------------------------|--|--|
| | „origin“ type of barrier | „buffer“ in meter (half width of real world object) | „likelihood“ being a barrier, with $0 \leq p \leq 1$ |
| streets | osm_street | highway = 'motorway': 5,25 highway != 'motorway': 3 | 1 |
| railroads | osm_railway | railway != 'tram': 3,75 railway = 'tram': 2,25 | 1 |
| waterways | osm_waterway | 1 | 1 |
| barriers | osm_barrier | 0 | 1 |
| trails | osm_trail | 1 | 0 ... 1 (from ground-truth) |
| land use change | lu_change | 0 | 0 ... 1 (from ground-truth) |

Fig. 19: additional attributes for line segments

- Second: polygon mesh (`ST_Polygonize()`) for different thresholds of „likelihood“ → representing different action spaces of activities

Polygon mesh

Selecting different polygons based on different intervals for barrier likelihood

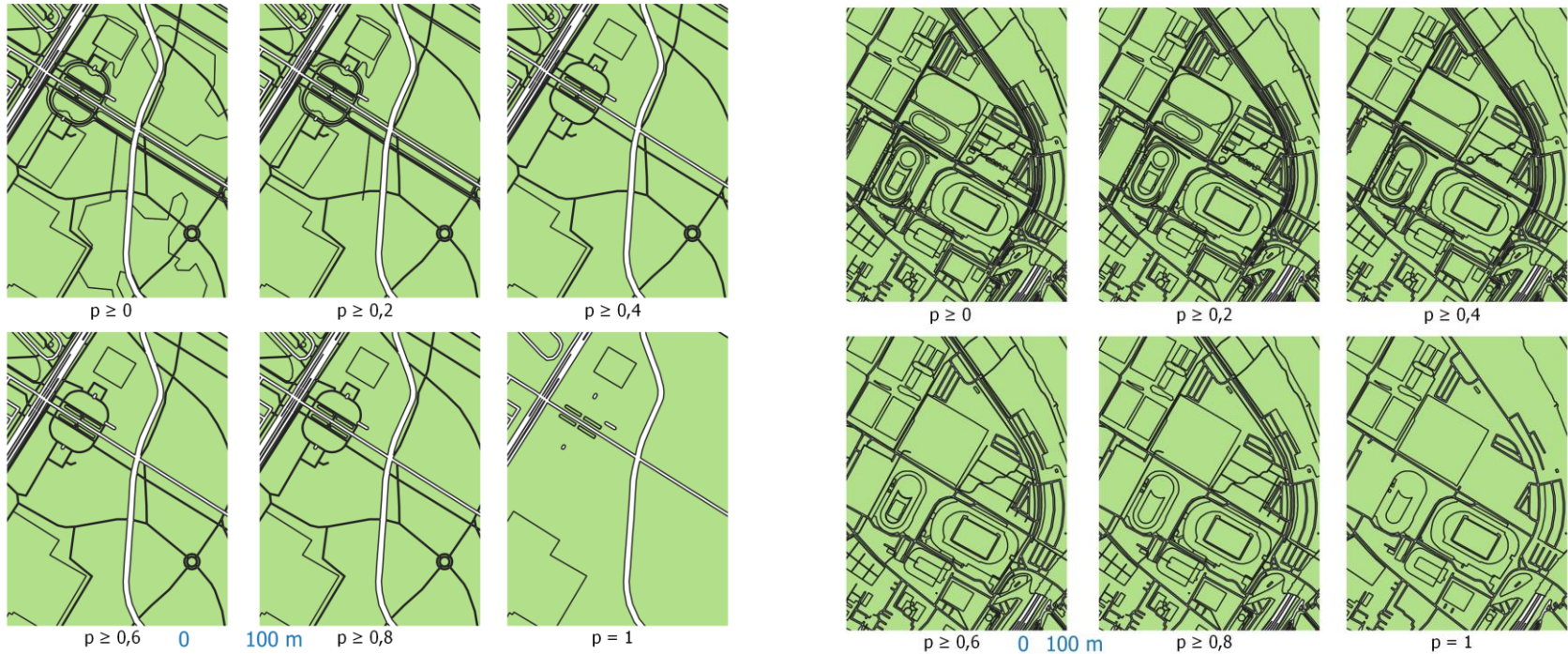


Fig. 20/21: polygon mesh in city park (left) and Ostragehege (right), Dresden

Model to predict publicly accessibility

Input features of each polygon

- Number of benches
- Number of waste baskets
- Number of public toilets
- Number of public internet / wifi

Reference data: cadastral parcels owned by the city of Dresden (17840 polygons)

Assumption: cadastral parcels owned by the city of Dresden will be publicly accessible

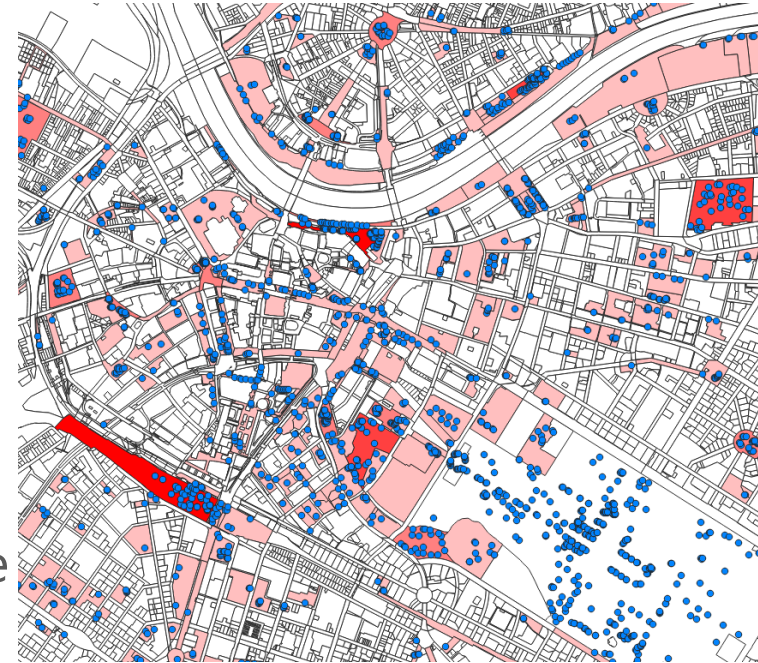


Fig. 22: relation polygons to benches

Logistic Regression

- Target
 - 0: not publicly accessible
 - 1: publicly accessible
- Counts (*in BBox*)
 - Total: 27126
 - 0: 26732
 - 1: 430

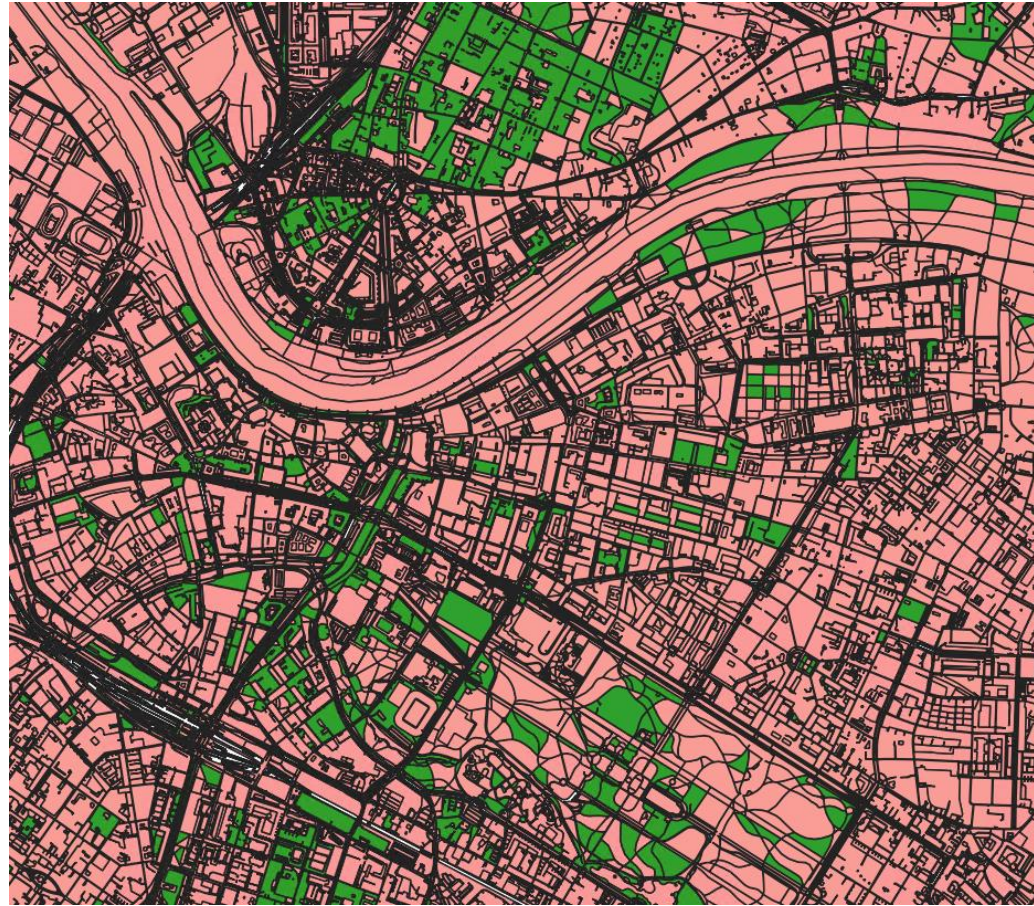


Fig. 23: result publicly accessibility

| bench (in_1) | waste basket (in_4) | public toilet (in_8) | Öffentliches Internet (in_9) | Score |
|---|------------------------|-------------------------|------------------------------------|----------|
| Logistic Regression | | | | |
| X | X | X | X | 0,767433 |
| X | X | X | | 0,767433 |
| X | X | | X | 0,767096 |
| X | | X | X | 0,766424 |
| | X | X | X | 0,767321 |
| Support Vector Classifier (SVC) | | | | |
| X | X | X | X | 0,767545 |
| X | X | X | | 0,767545 |
| X | X | | X | 0,767321 |
| X | | X | X | 0,766200 |
| | X | X | X | 0,767321 |
| SVC with Radial Basis Function (C= 1E6, gamma= 1.) | | | | |
| X | X | X | X | 0,772197 |
| X | X | X | | 0,771804 |
| X | X | | X | 0,771637 |
| X | | X | X | 0,770123 |
| | X | X | X | 0,769170 |

Fig. 24: intrinsic score

Model to predict greenness

Input features of each polygon

- Number of benches
- Number of picnic tables
- Number of trees
- Number of waste baskets

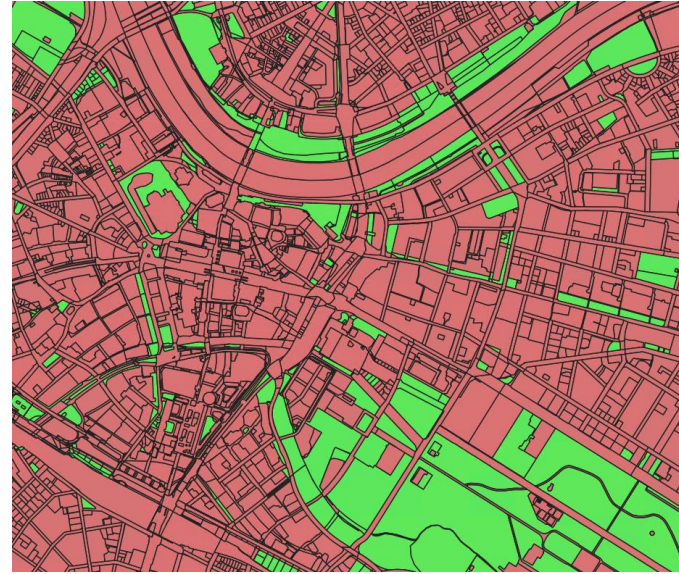


Fig. 25: reference data showing greenness

Reference data: official cadastral data (ALKIS) + land use information (22753 polygons)

Assumption: land use type „Wald“, „Gehölz“, „Friedhof“, „Sport-, Freizeit- und Erholungsfläche“ represents greenness

Logistic Regression

- Target
 - 0: not green
 - 1: green
- Counts (*in BBox*)
 - Total: 26472
 - 0: 26732
 - 1: 690



Fig. 26: result greenness

Model to predict greenness

| bench (in_1) | picnic table (in_2) | tree (in_3) | waste baskets (in_4) | Score |
|---|------------------------|----------------|-------------------------|----------|
| Logistic Regression | | | | |
| X | X | X | X | 0,922867 |
| X | X | X | | 0,922779 |
| X | X | | X | 0,922867 |
| X | | X | X | 0,922867 |
| | X | X | X | 0,922691 |
| Support Vector Classifier (SVC) with linear Kernel and C=1 | | | | |
| X | X | X | X | 0,926018 |
| X | X | X | | 0,925432 |
| X | X | | X | 0,926018 |
| X | | X | X | 0,926018 |
| | X | X | X | 0,926018 |

Fig. 27: intrinsic score

Conclusion and outlook

- A new approach of generating urban green spaces
- First testing show good results

Outlook

- E. g.: XGBoost, grid search for parameters, feature importance
- Generate further input features (path density, other POIs, geometry...)
- Intersect with greenness from remote sensing
- Testing in further cities, mapping more barrier probabilities

Return to OpenStreetMap project

- Completeness analysis of barrier=* ?
- Creating an assistant layer as a help to map land use / land cover ?
- Knowledge about barrier probability and publicly accessibility could be useful to improve routing, e.g. Open Space routing (through polygons)
- ...



Thank you for your attention!

Theodor Rieche
t.rieche@ioer.de

source code & master thesis:

<https://github.com/traveller195/master-thesis-green-spaces-derived-from-osm>