

# SA-TIED Geospatial Analysis Workshop

## Overview



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# Objectives

We will explore the following topics:

- Fundamentals of using R for data analysis.
- Creating thematic maps using R.
- Quantifying the degree of spatial dependence in a dataset.
- Incorporating space into statistical models.

# Schedule

Day 1 – Morning	R for Data Analysis
Day 1 – Afternoon	R for Spatial Analysis
Day 2 – Morning	Spatial Autocorrelation
Day 2 – Afternoon	Spatial Models



# SA-TIED Geospatial Analysis Workshop

## S04 – Spatial Models



# This session

- Spatial models.
- Geographically weighted statistics.
- Geographically weighted associations.
- More spatial analysis.

# Linear models

- Used to determine the relationship or association between a dependent with one or more independent variables.
- Important assumptions: homoscedasticity and independence of residuals.
- Violating this assumption can lead to inefficient estimates and unreliable hypothesis tests.

# Linear models

When building a model based on spatial data:

- Map the residuals of the linear model to visually inspect for spatial patterns.
- Calculate Moran's I statistic on the residuals to assess spatial autocorrelation.
- If spatial autocorrelation is present, fit a **spatial linear model** to account for it.
- Recalculate Moran's I statistic on the residuals of the spatial model to confirm that the autocorrelation has been addressed.

# Spatial models

A **spatial error model** adjusts for spatial autocorrelation by adding a spatially lagged error term to the regression equation:

$$y = X\beta + v, v = \lambda Wv + \epsilon$$

where  $X\beta$  represents the standard regression components,  $\lambda$  is a spatial autoregressive parameter,  $W$  represents the spatially weights matrix, and  $u$  is a vector of spatially autocorrelated errors.



# Spatial models

A **spatial lag model** incorporates a spatially lagged dependent variable, which is the weighted sum of the dependent variable values in neighboring locations, into the regression equation:

$$\mathbf{y} = \rho \mathbf{W}\mathbf{y} + \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$$

where  $\rho$  is the spatial autoregressive coefficient,  $\mathbf{W}\mathbf{y}$  represents the spatially lagged dependent variable, and  $\mathbf{X}\boldsymbol{\beta}$  represents the standard regression components.

# Spatial models

- Both the spatial error and spatial lag models assume that the relationships between variables are the same across the study area, with adjustments made only for spatial dependencies.
- A Lagrange Multiplier Test can be used to make a decision as to which of these two models is most appropriate.
- What about non-stationarity?

# Geographically weighted statistics

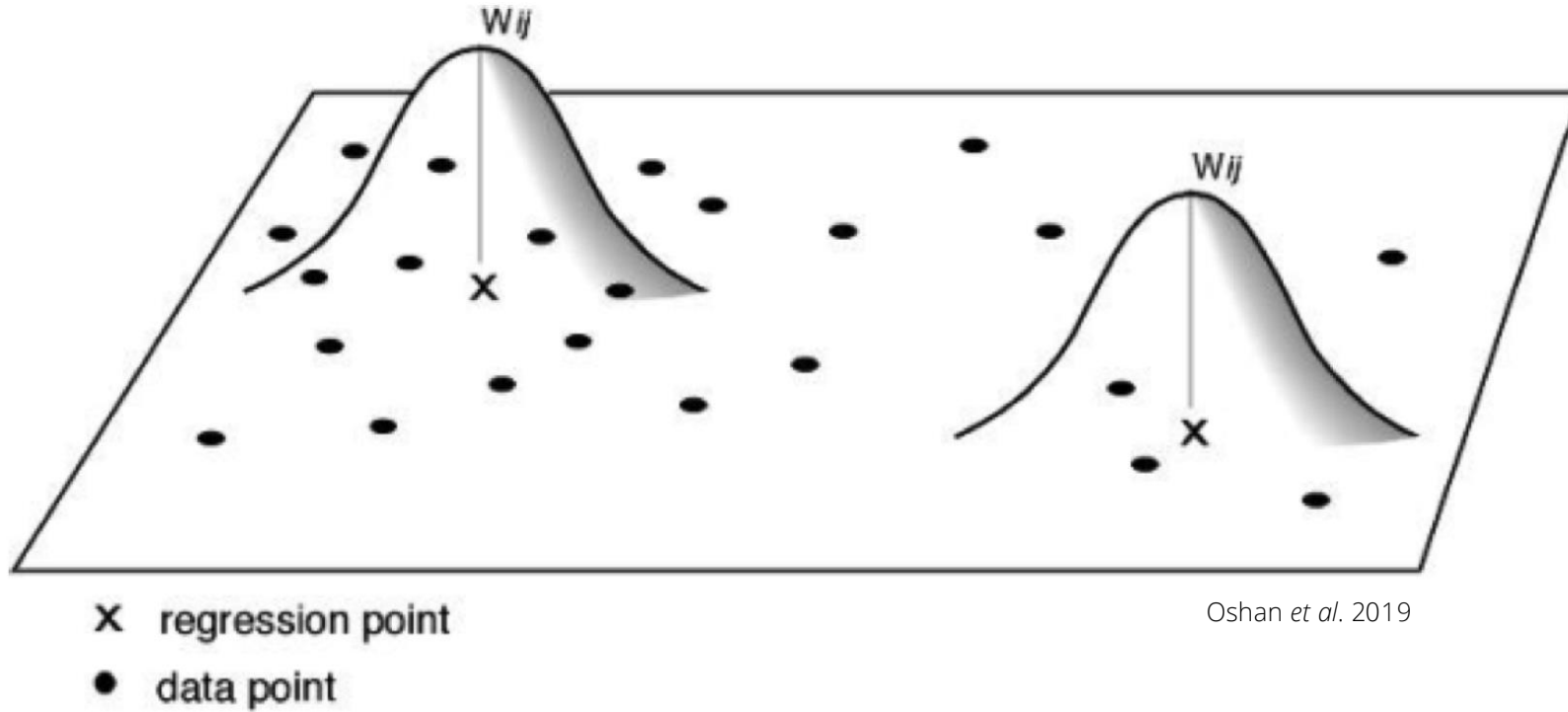
- Unlike traditional global models, which estimate a single set of parameters for the entire study area, geographically weighted statistics allow for parameter estimates that vary across different locations.
- Local means, local standard deviations, local variances.
- Typically uses some **kernel function** to weigh observations based on their distance from the location of interest.

# Geographically weighted statistics

“Everything is related to everything else, but near things are more related than distant things.”

Walter Tobler 1970

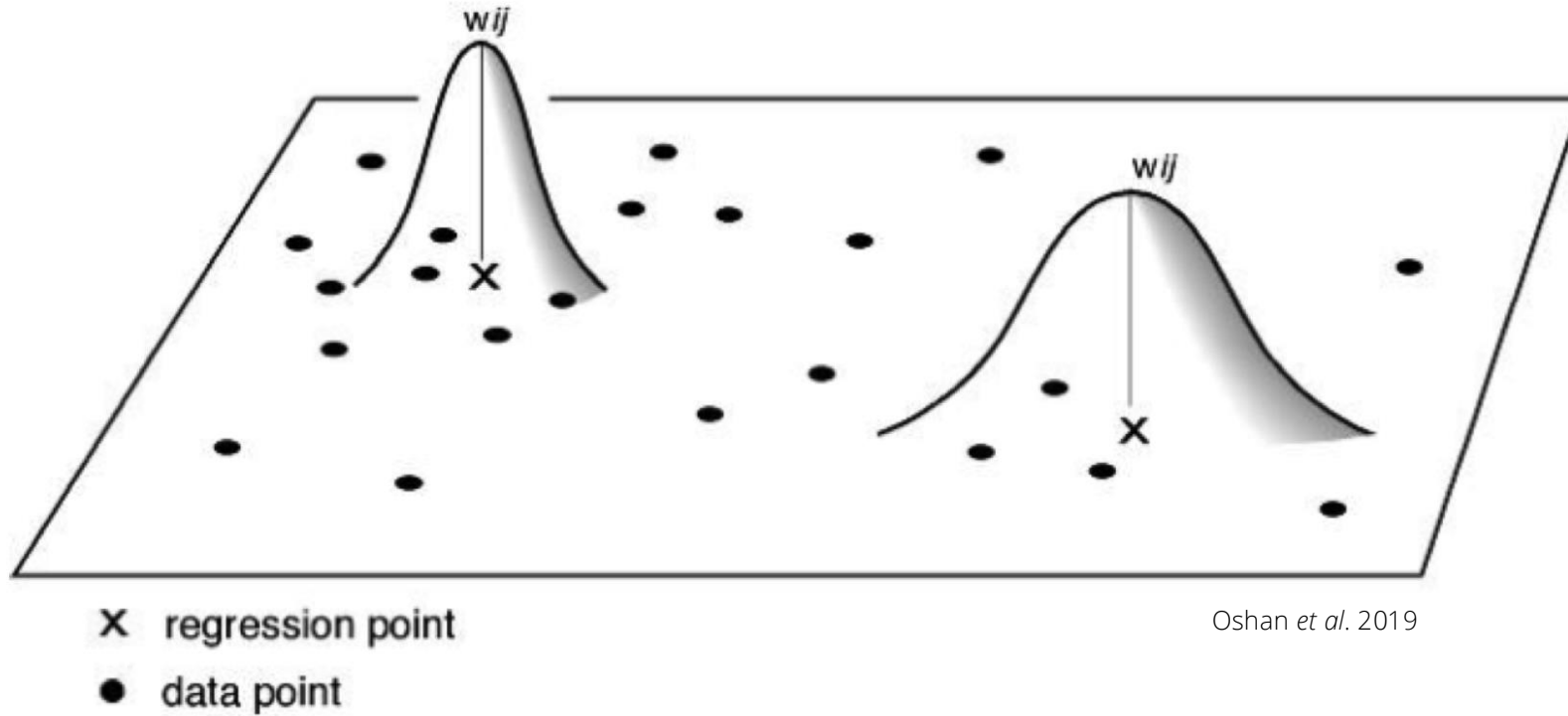
# Geographically weighted statistics



Oshan *et al.* 2019



# Geographically weighted statistics



Oshan *et al.* 2019

# Geographically weighted associations

- These ideas can be extended to correlation and regression:
  - Geographically weighted correlation (GWC)
  - Geographically weighted regression (GWR)
- The basic GWR equation is:

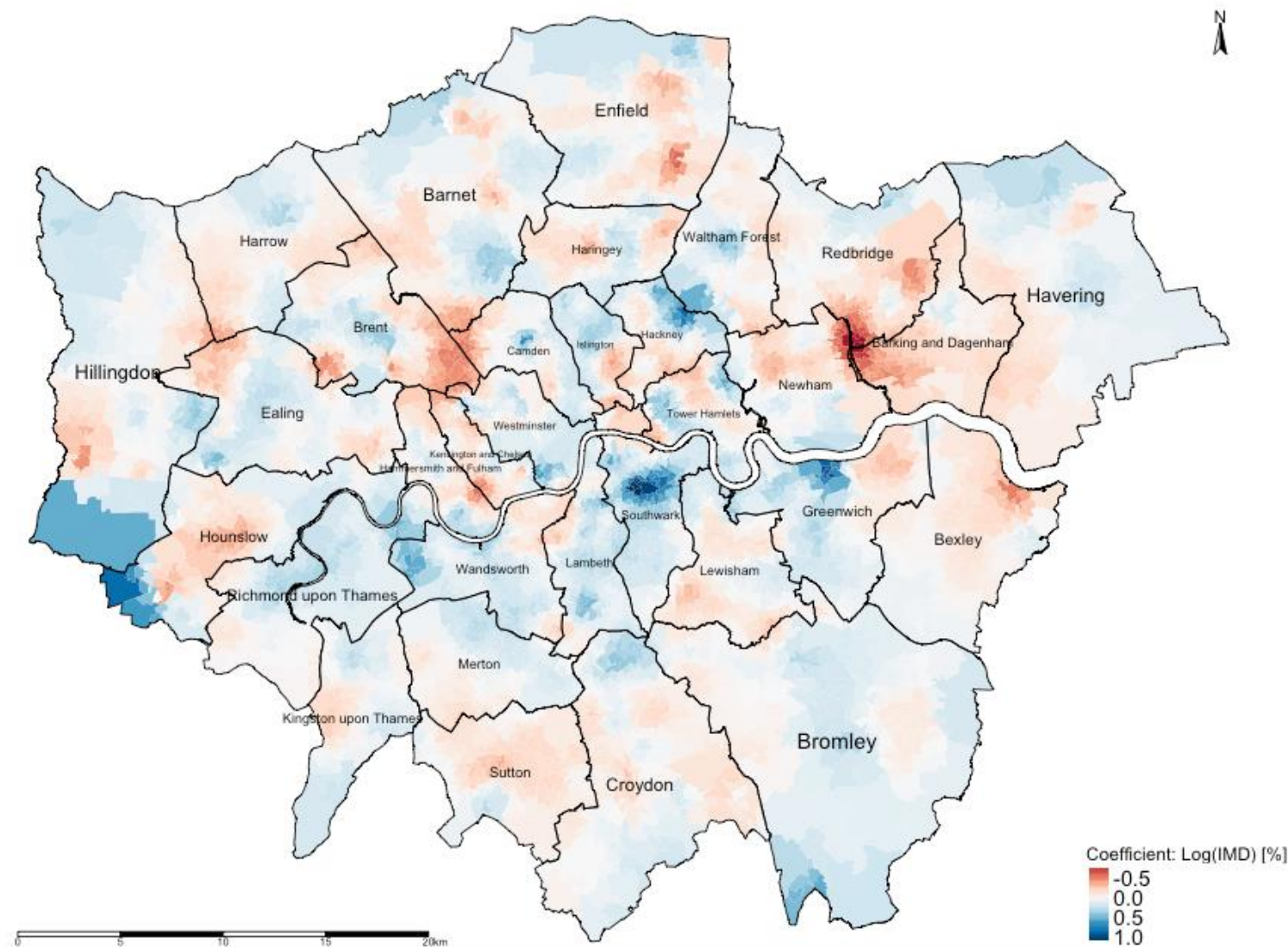
$$y_i = \beta_0(v_i, v_i) + \sum_{k=1}^p \beta_k(v_i, v_i) x_{ik} + \epsilon_i$$

where  $(v_i, v_i)$  are the coordinates of location  $i$  and  $\beta_k(v_i, v_i)$  are the location-specific coefficients.

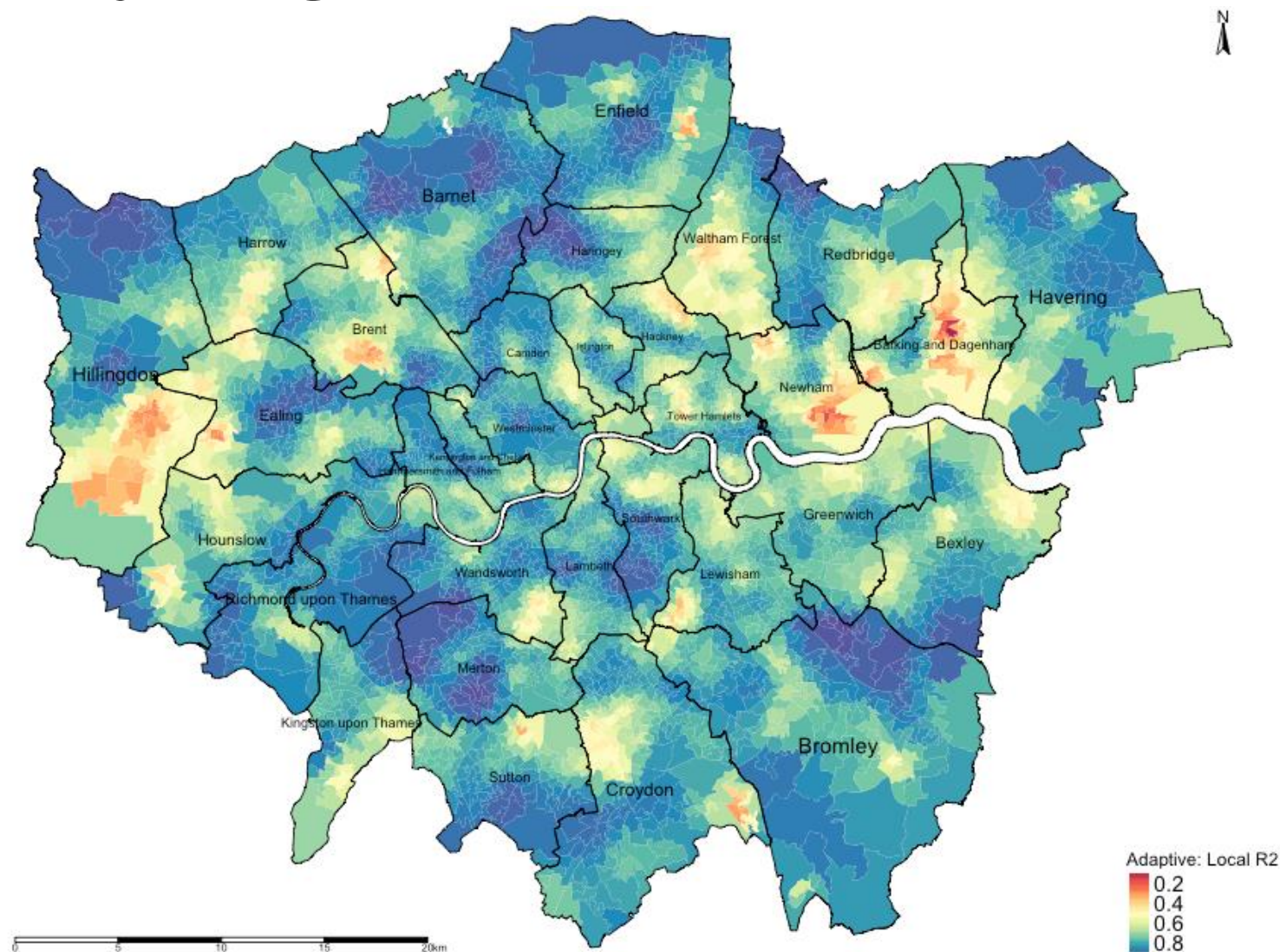
# Geographically weighted associations

- Each area has its own set of regression coefficients.
- Each location has its own  $R^2$  value.
- Each area has its own standard errors for the coefficient.
- More recently: bandwidths can vary between different variables.

# Geographically weighted associations



# Geographically weighted associations





# Questions

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# SA-TIED Geospatial Analysis Workshop

More spatial analysis



# More spatial analysis

- File formats.
- Map projections.
- Digitisation and geocoding.
- Spatial operations.
- Accessibility analysis.
- Geodemographic classification.
- Raster data.

# File formats: Geopackage

- A GeoPackage is an open, standards-based, platform-independent, portable, self-describing, compact format for transferring geospatial data.
- It stores spatial data layer as a single file, based upon an SQLite database.
- How to spot in the wild: `.gpkg`

# File formats: Shapefile

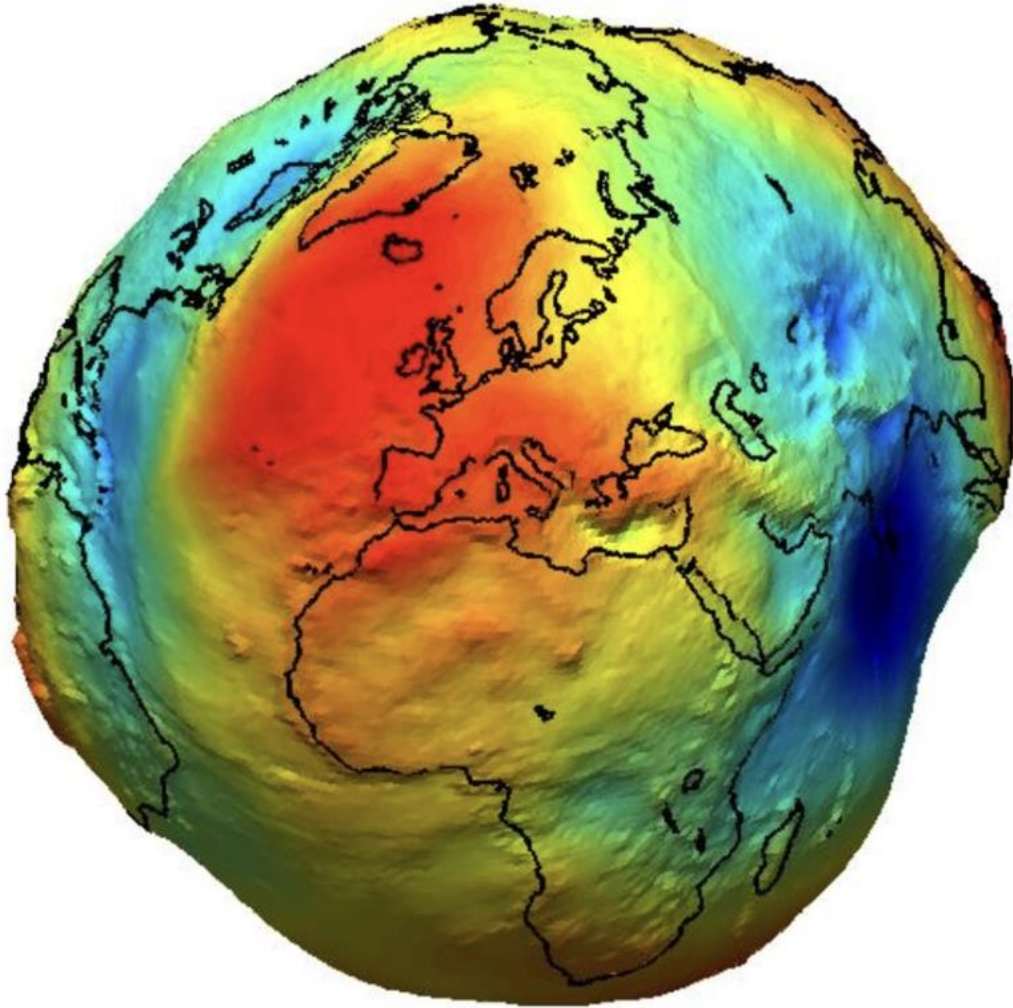
- `.shp` contains the feature geometry. *Mandatory.*
- `.shx` index file which stores the position of the feature's ID in the `.shp` file. *Mandatory.*
- `.dbf` stores all attribute information associated with the records. *Mandatory.*
- `.prj` contains the coordinate system information and projection. *Optional but not really.*
- `.xml` general metadata. *Optional.*
- `.cpg` encoding information. *Optional.*
- `.sbn` optimisation file for spatial queries. *Optional.*



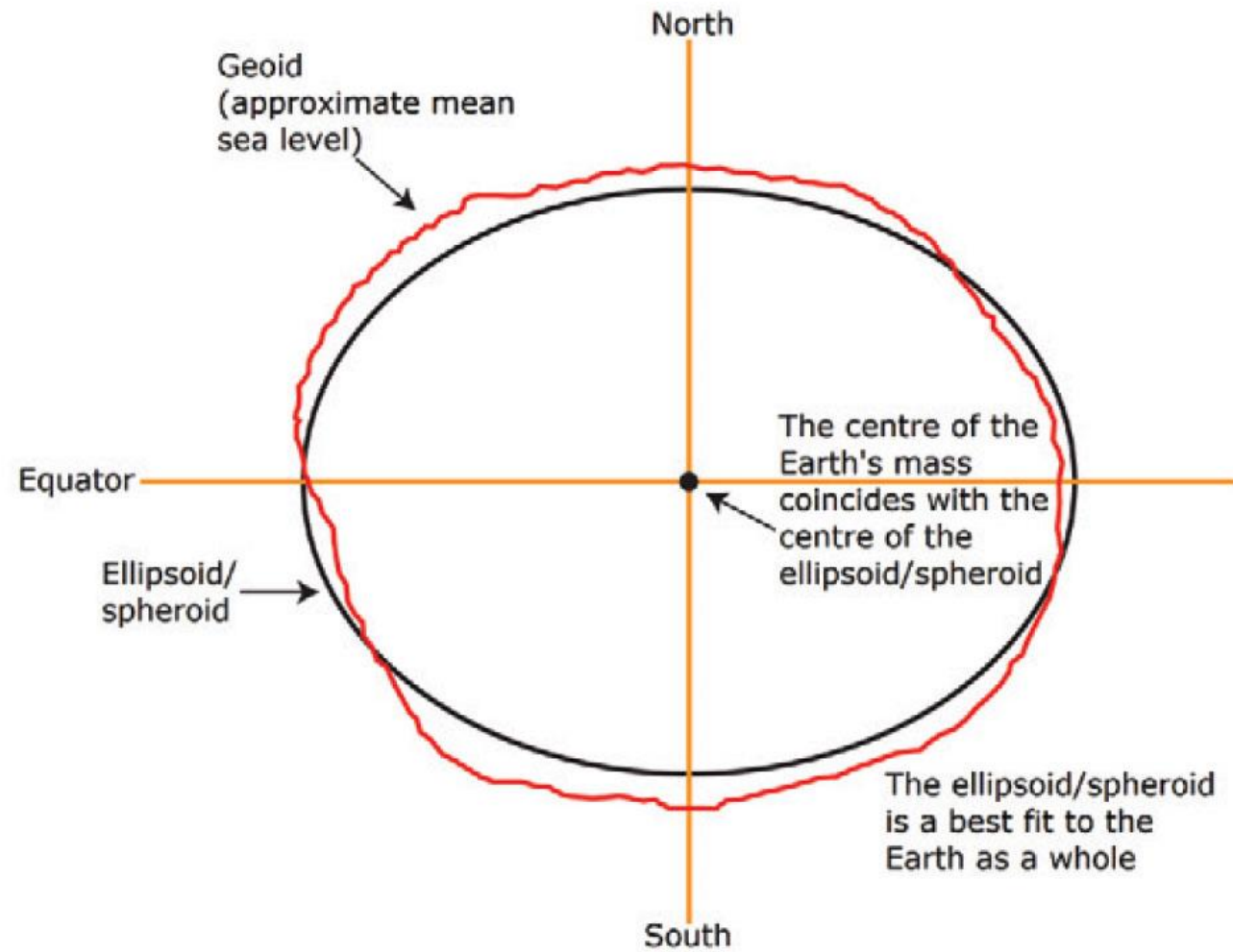
# File formats: Shapefile



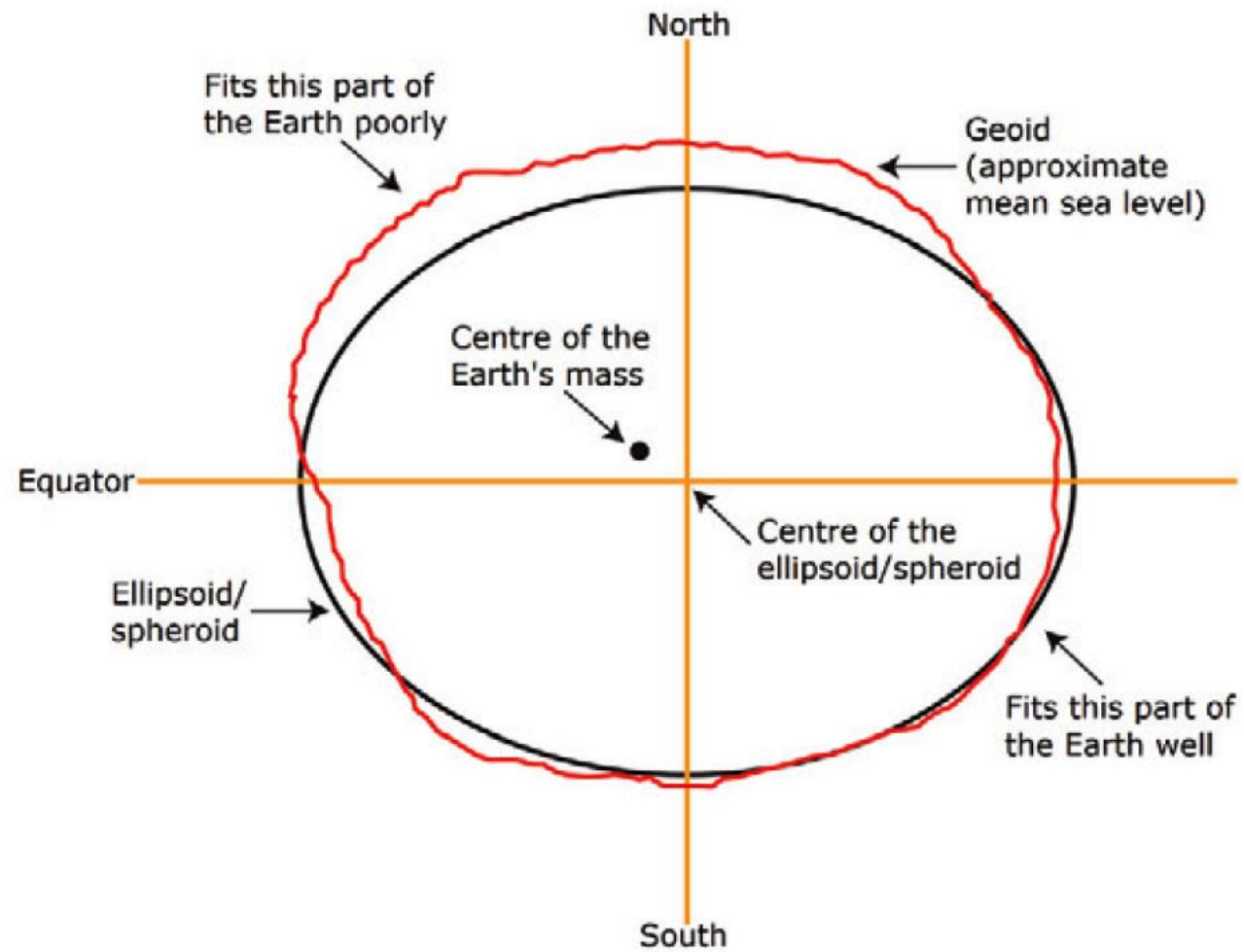
# Map projections



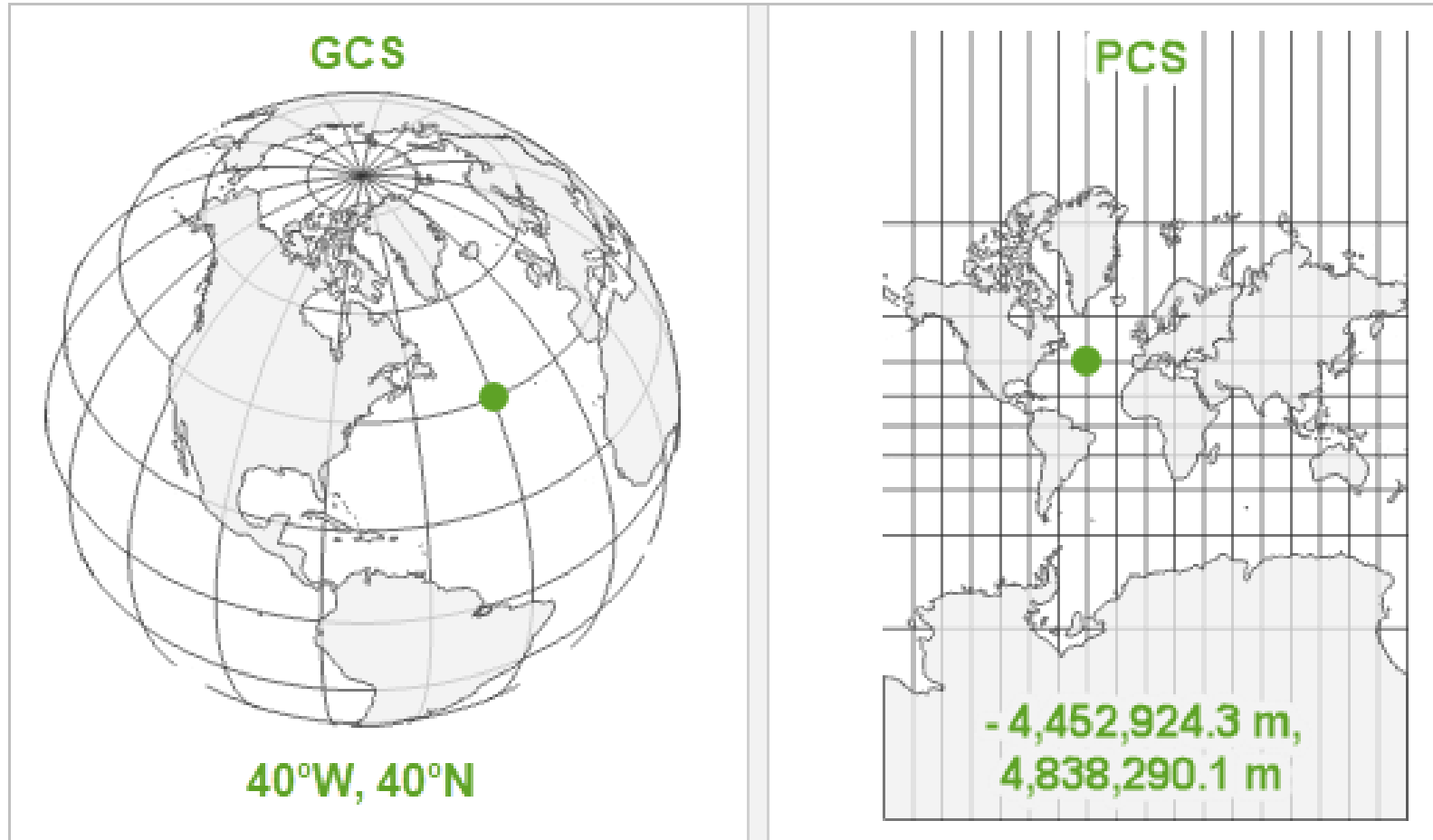
# Map projections



# Map projections



# Map projections

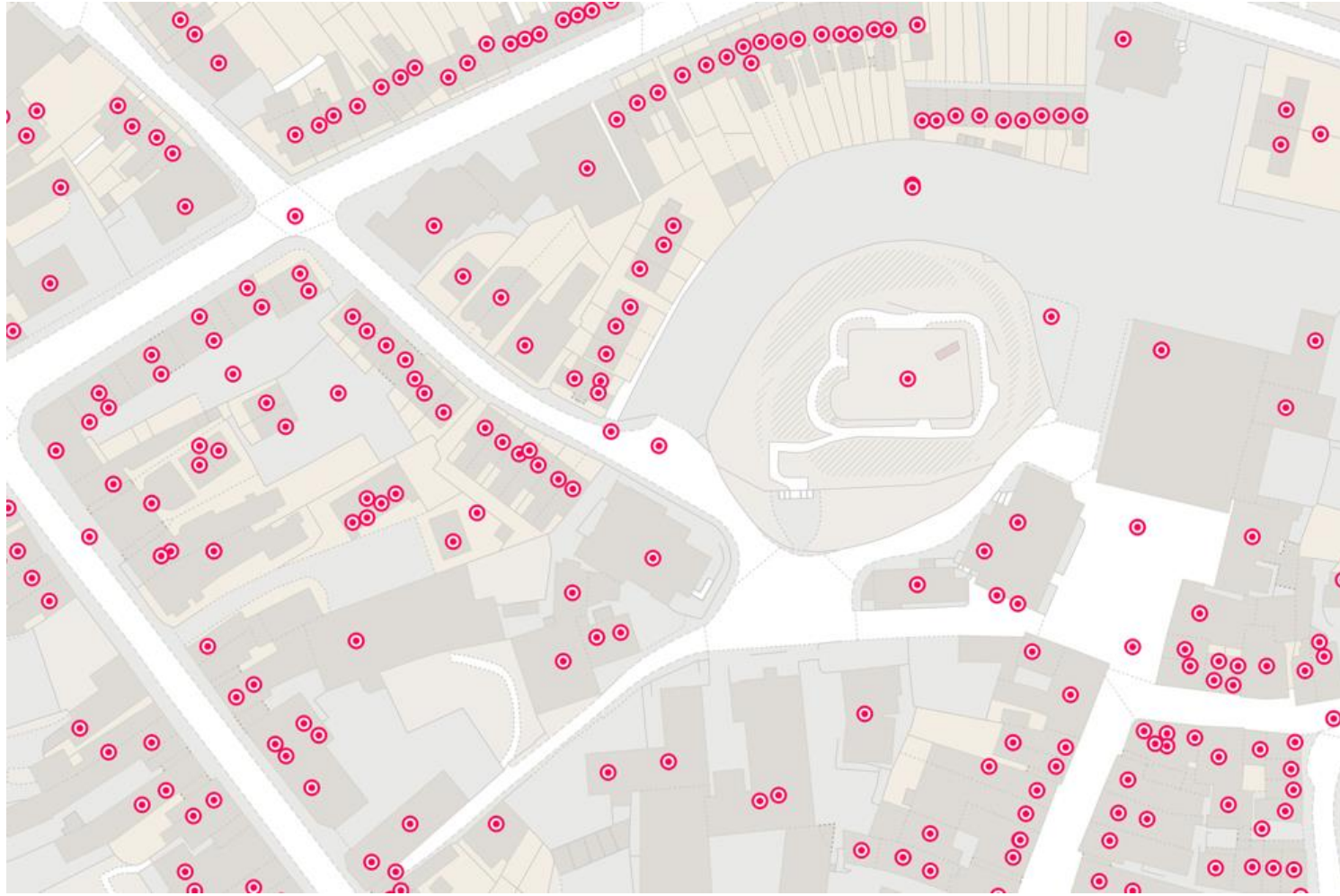




# Geocoding

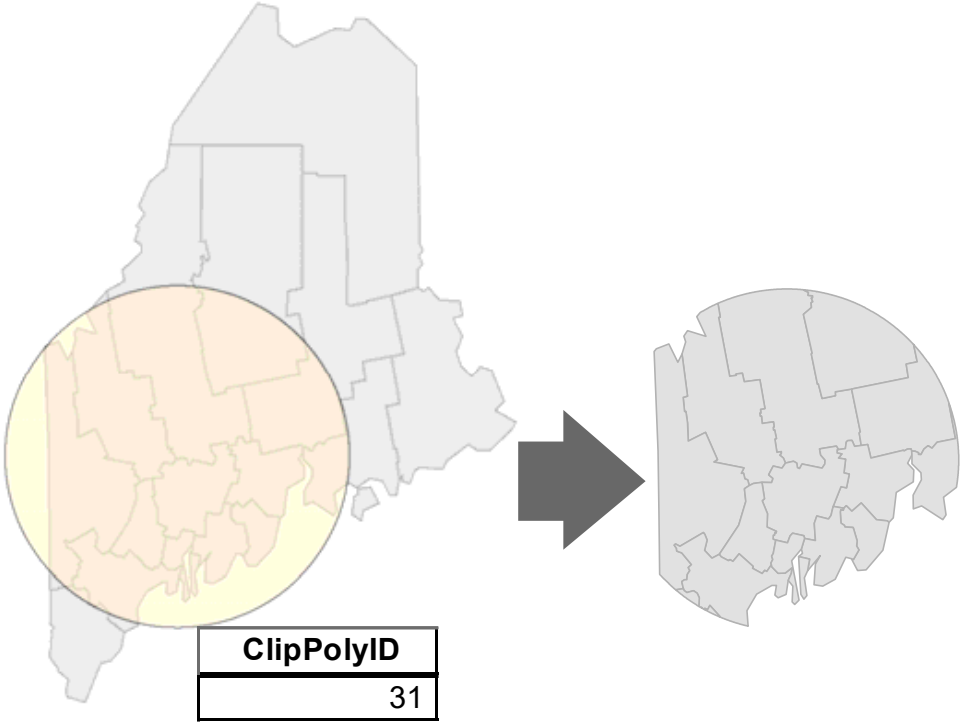
Forename	Surname	Address
Justin	van Dijk	Flat 18 Terry House SW22NT London

# Geocoding



# Spatial operations

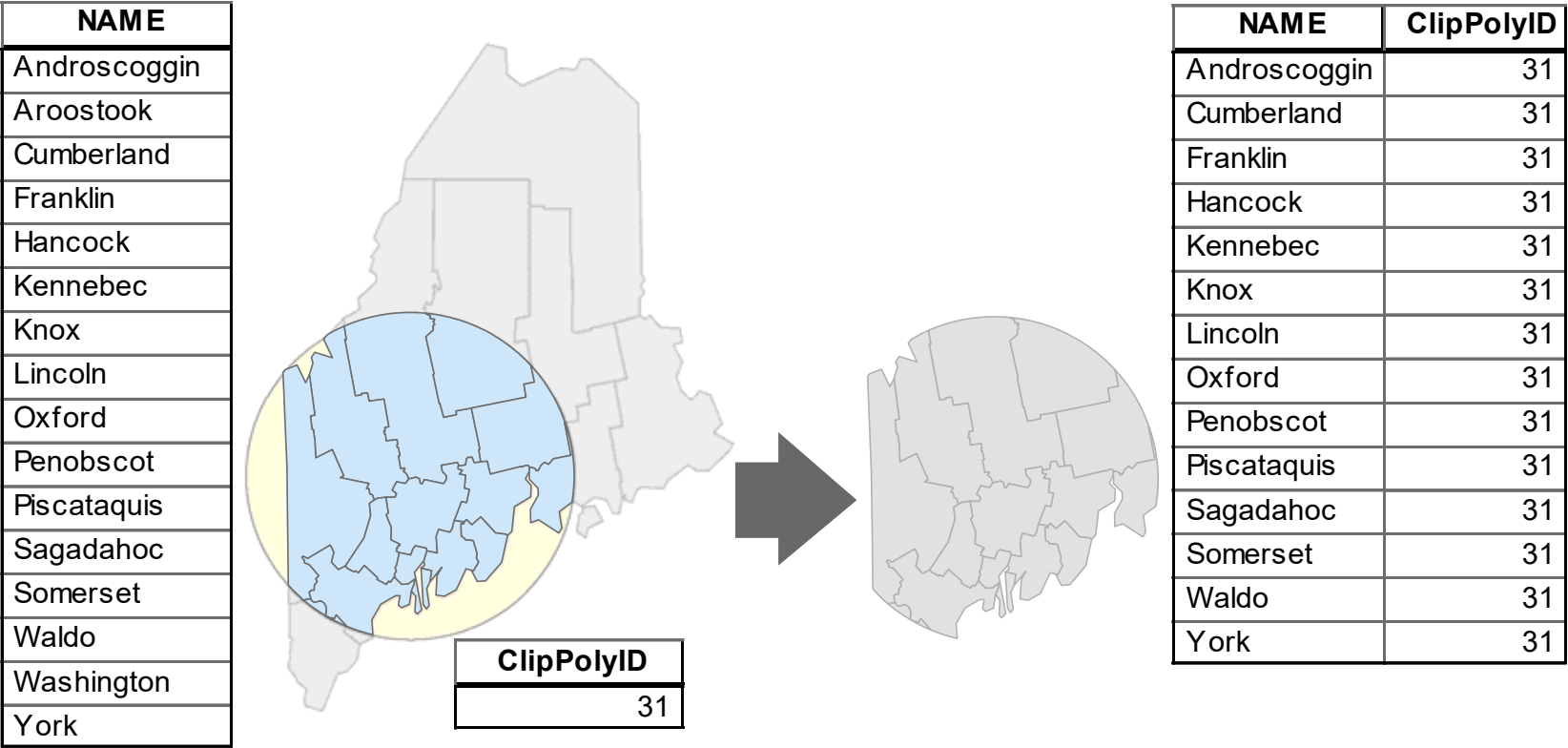
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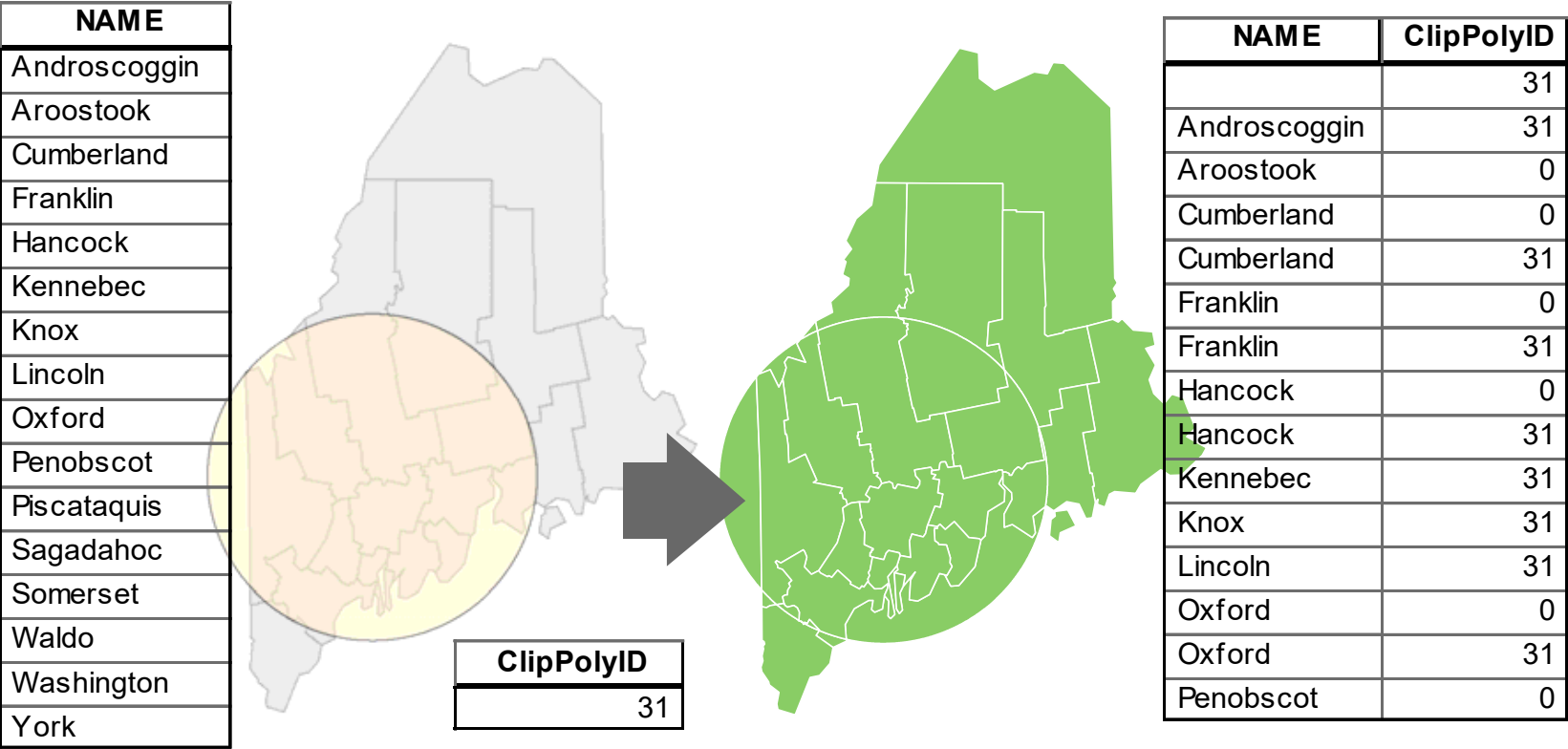
Gimond, M. 2021. Intro to GIS and Spatial Analysis. [online]  
<https://mgimond.github.io/Spatial/introGIS.html>

# Spatial operations



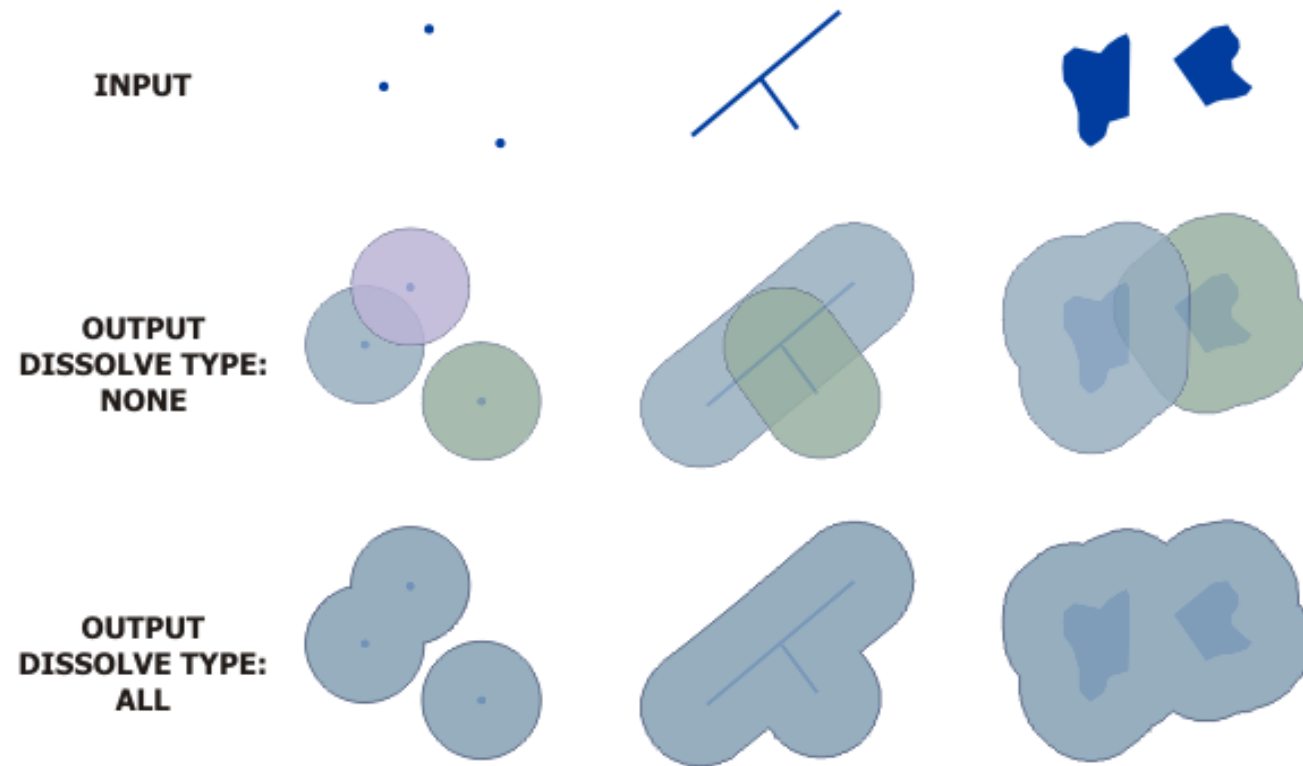
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# Spatial operations



Gimdond, M. 2021. Intro to GIS and Spatial Analysis. [online]  
<https://mgimond.github.io/Spatial/introGIS.html>

# Spatial operations



ESRI. 2021. Buffer. [online]

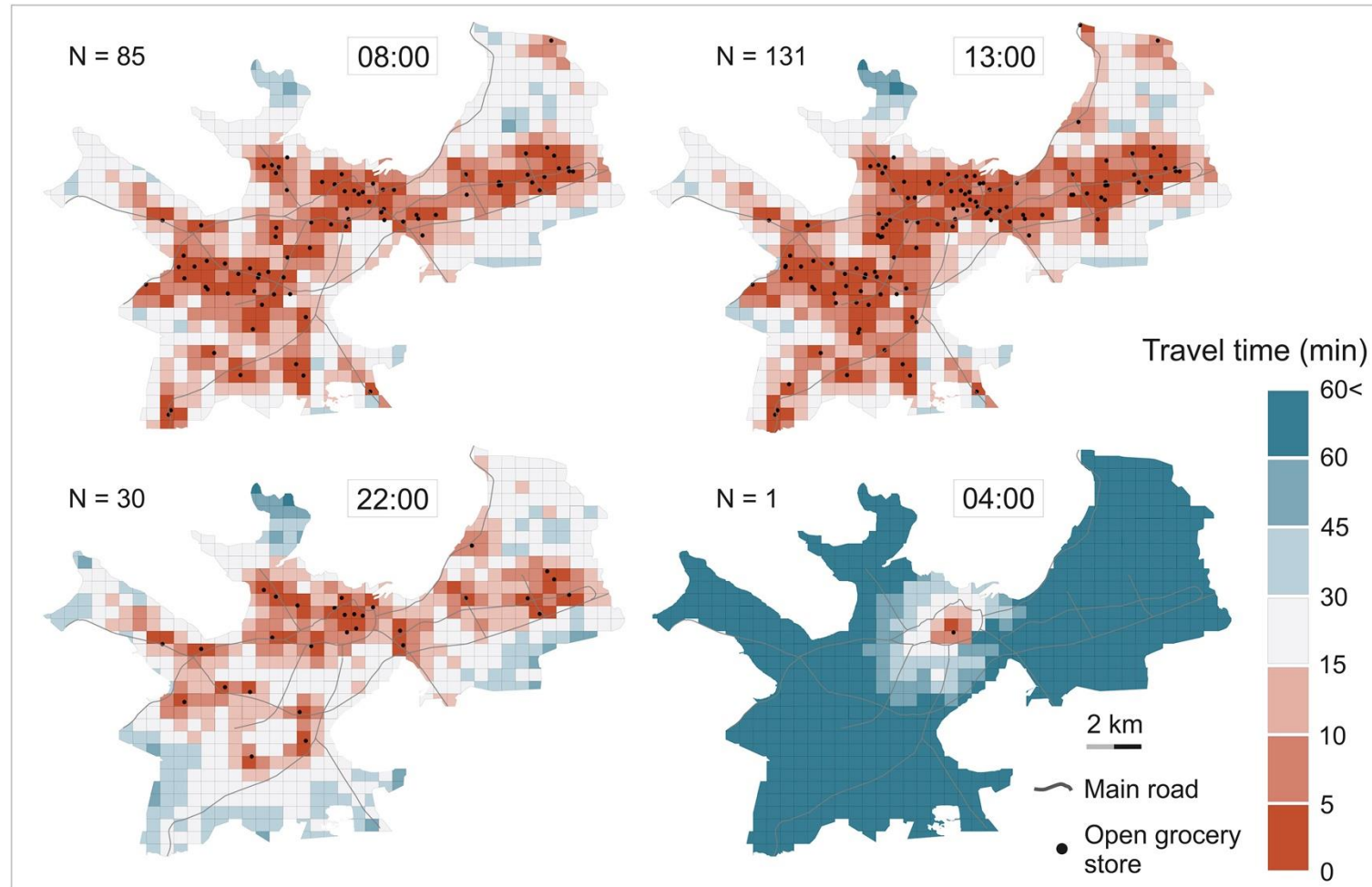
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# Spatial operations





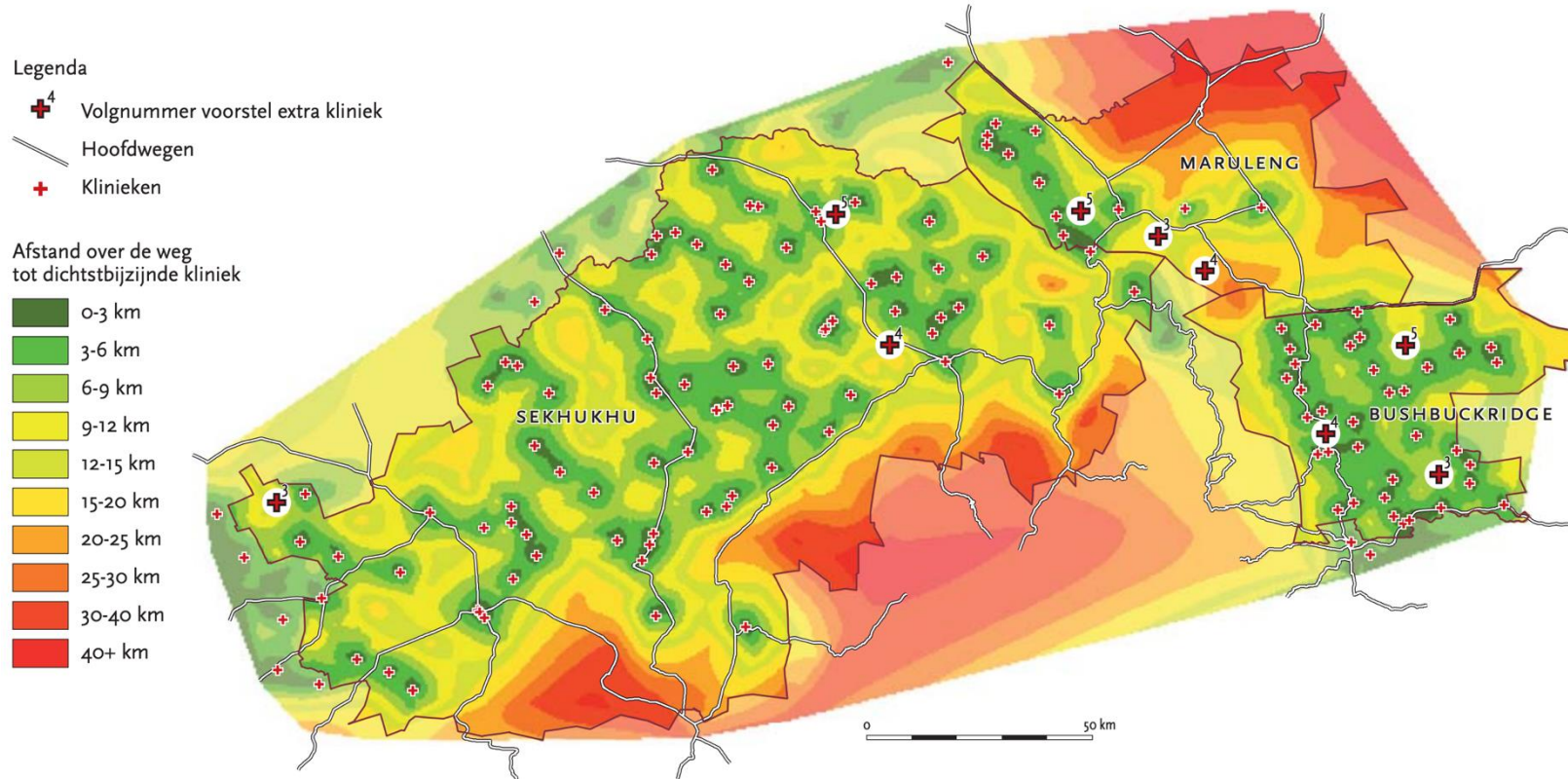
# Accessibility analysis



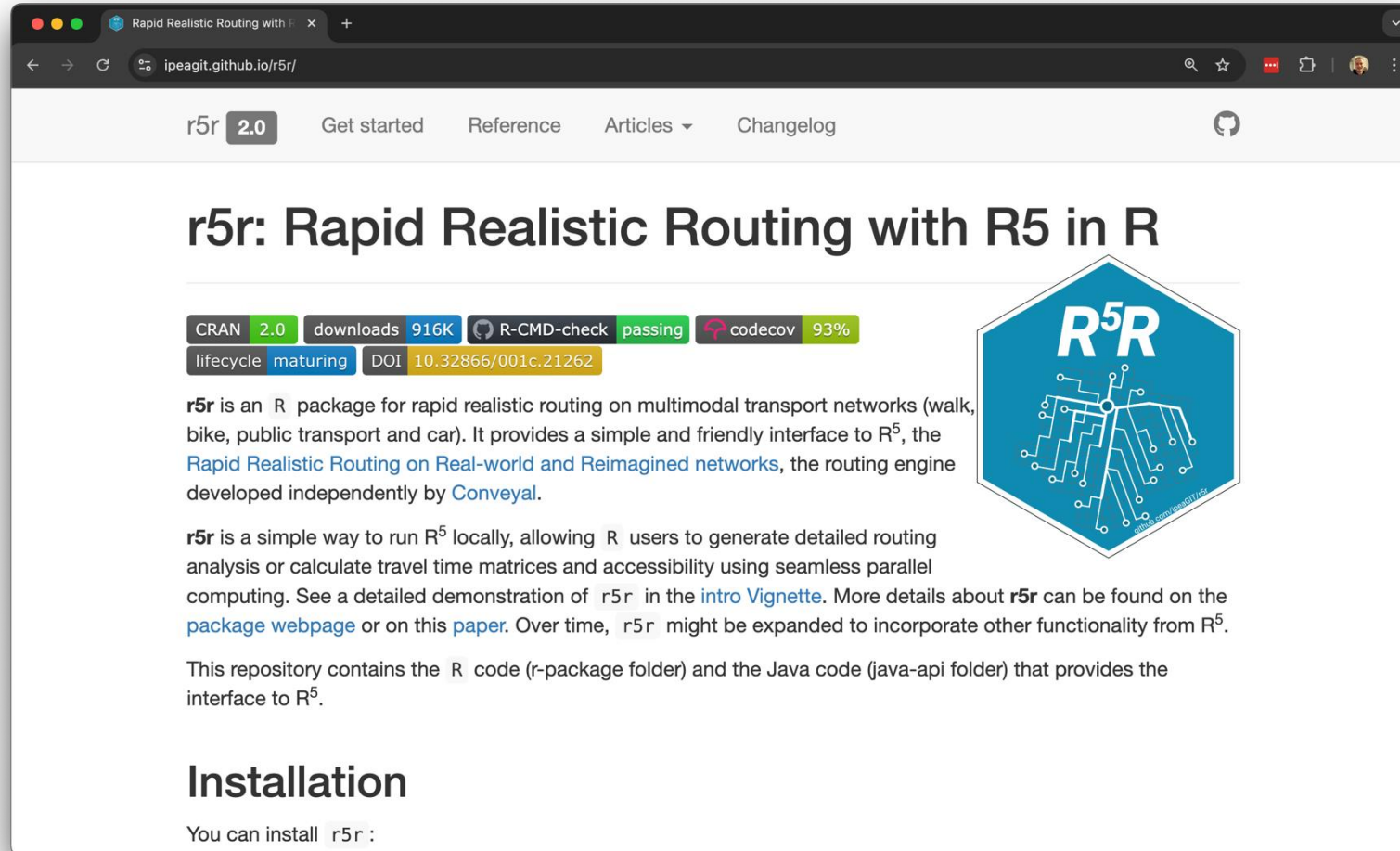
# Accessibility analysis



# Accessibility analysis



# Accessibility analysis



The screenshot shows the GitHub repository page for **r5r** (Rapid Realistic Routing with R5 in R) by ipeagit. The page features a navigation bar with links to 'Get started', 'Reference', 'Articles', and 'Changelog'. The main heading is 'r5r: Rapid Realistic Routing with R5 in R'. Below this, a row of badges displays project statistics: CRAN version 2.0, 916K downloads, R-CMD-check passing, codecov 93%, lifecycle maturing, and DOI 10.32866/001c.21262. A blue hexagonal logo with 'R<sup>5</sup>R' and a circuit-like pattern is positioned to the right. The text describes **r5r** as an R package for multimodal transport routing, providing an interface to R<sup>5</sup>. It mentions that **r5r** allows for local execution and detailed routing analysis, and that the repository includes both R and Java code.

**r5r** 2.0 Get started Reference Articles Changelog

## r5r: Rapid Realistic Routing with R5 in R

CRAN 2.0 downloads 916K R-CMD-check passing codecov 93% lifecycle maturing DOI 10.32866/001c.21262

**r5r** is an R package for rapid realistic routing on multimodal transport networks (walk, bike, public transport and car). It provides a simple and friendly interface to R<sup>5</sup>, the [Rapid Realistic Routing on Real-world and Reimagined networks](#), the routing engine developed independently by [Conveyal](#).

**r5r** is a simple way to run R<sup>5</sup> locally, allowing R users to generate detailed routing analysis or calculate travel time matrices and accessibility using seamless parallel computing. See a detailed demonstration of **r5r** in the [intro Vignette](#). More details about **r5r** can be found on the [package webpage](#) or on this [paper](#). Over time, **r5r** might be expanded to incorporate other functionality from R<sup>5</sup>.

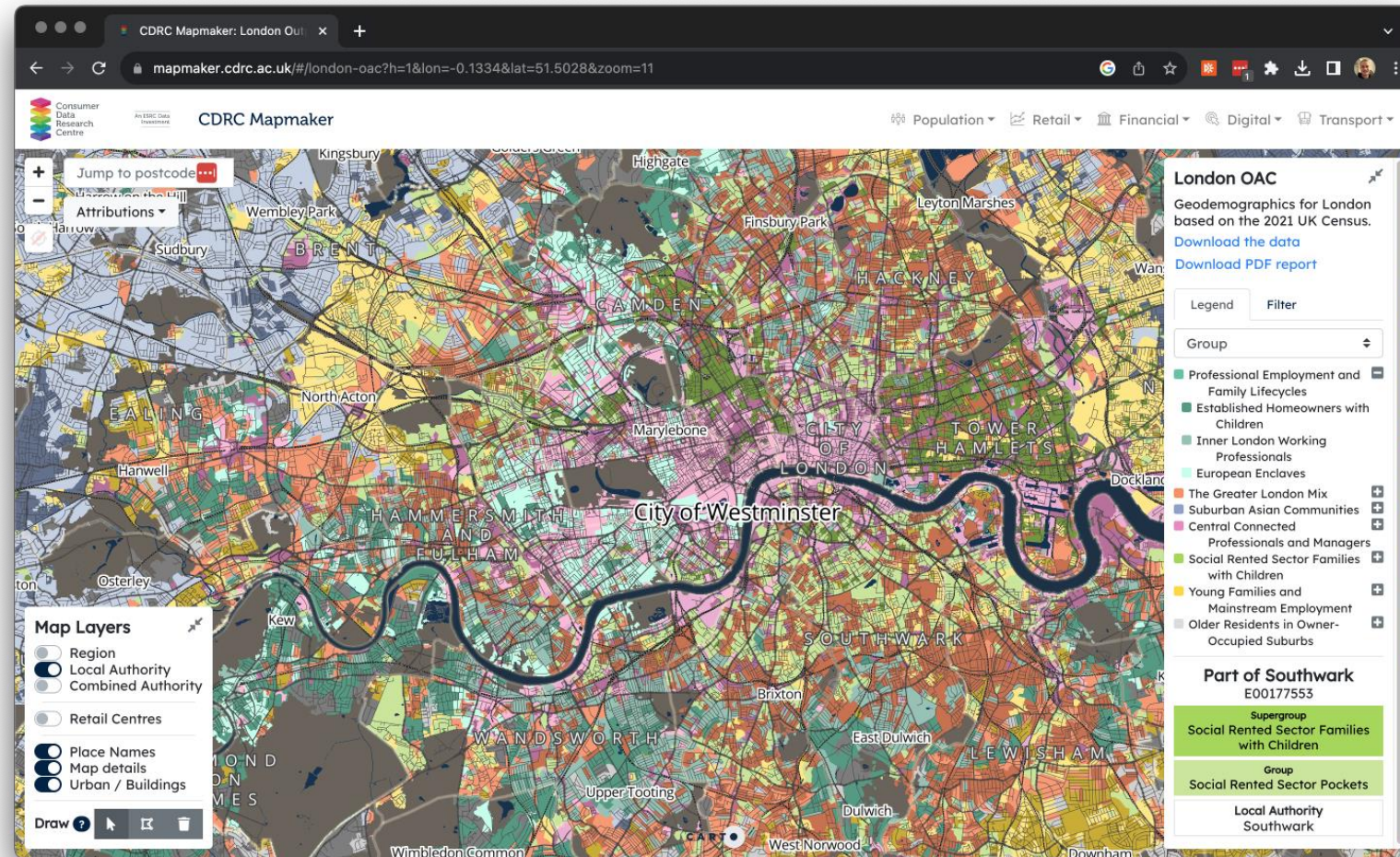
This repository contains the R code (r-package folder) and the Java code (java-api folder) that provides the interface to R<sup>5</sup>.

## Installation

You can install **r5r** :



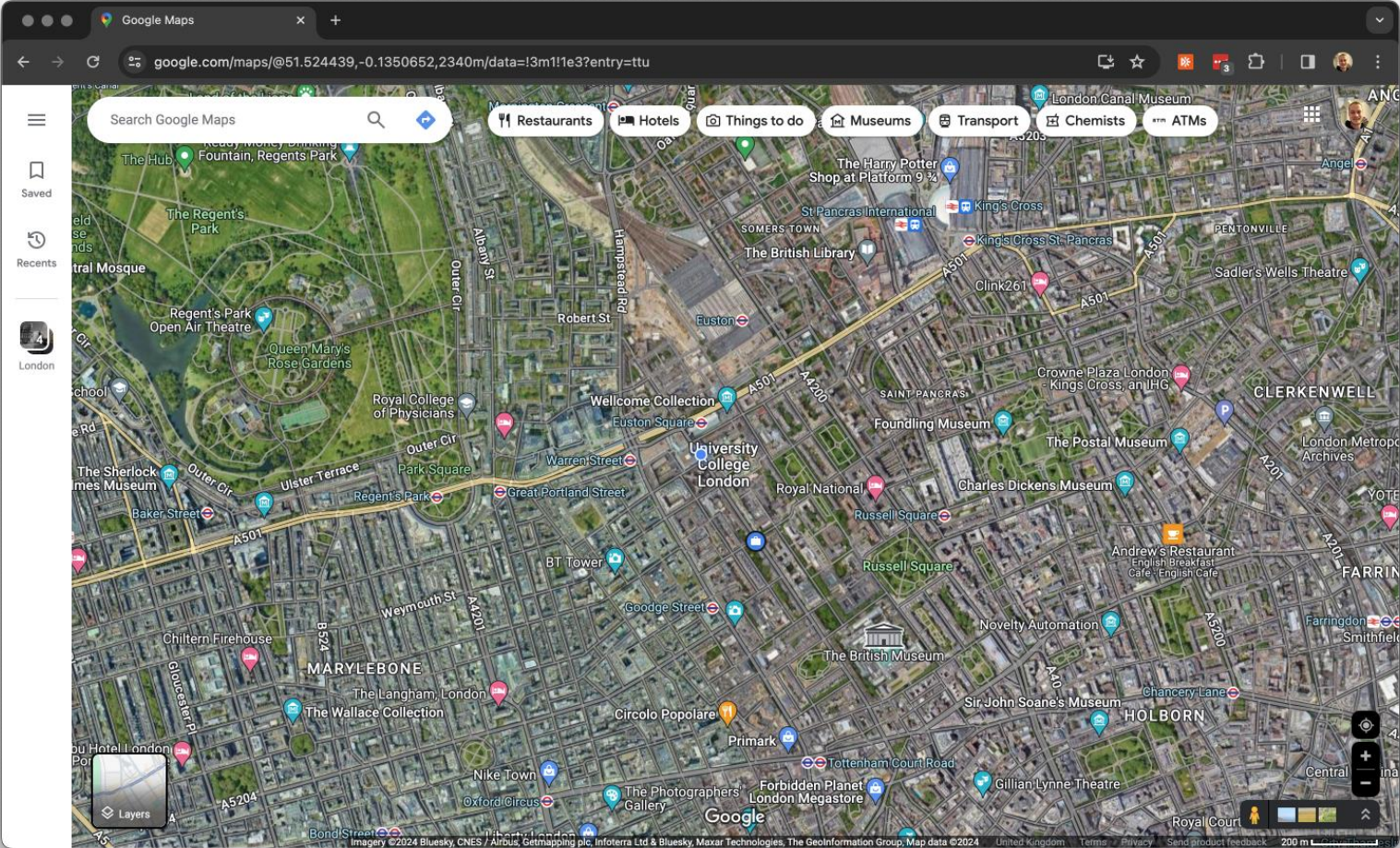
# Geodemographic classification



2021 London Output Area Classification on [mapmaker.cdrc.ac.uk](http://mapmaker.cdrc.ac.uk)

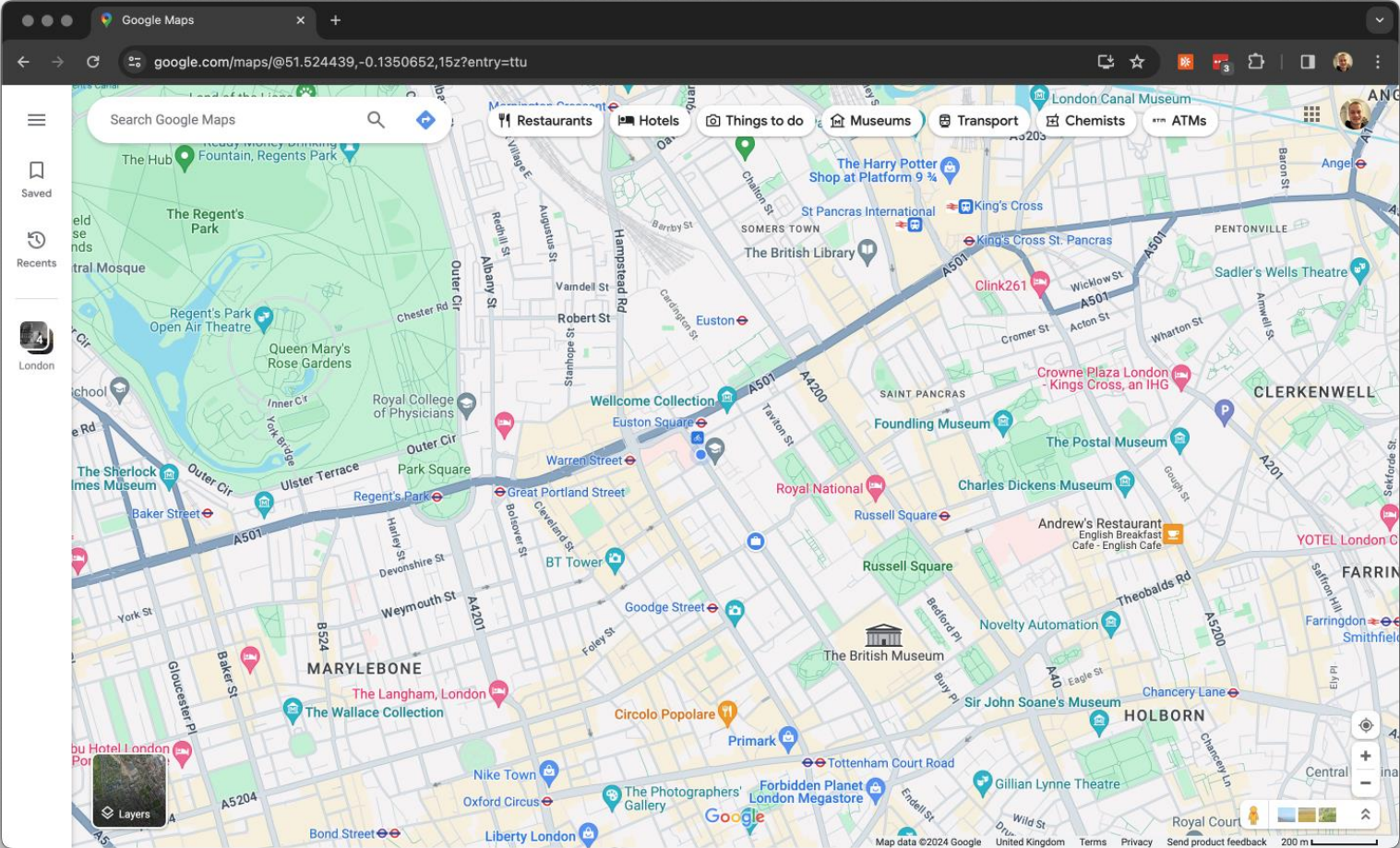


# GIScience





# GIScience





# GIScience

- GIScience relies on representing spatial information in a digital format. Traditionally, geographic information is conveyed in two primary ways:.

**Vector** This method uses a finite set of discrete geometric objects, such as points, lines, and polygons, to represent spatial features.

**Raster** This approach employs images or grids to represent surfaces, with each cell or pixel holding a value, often indicating attributes like colors or measurements.



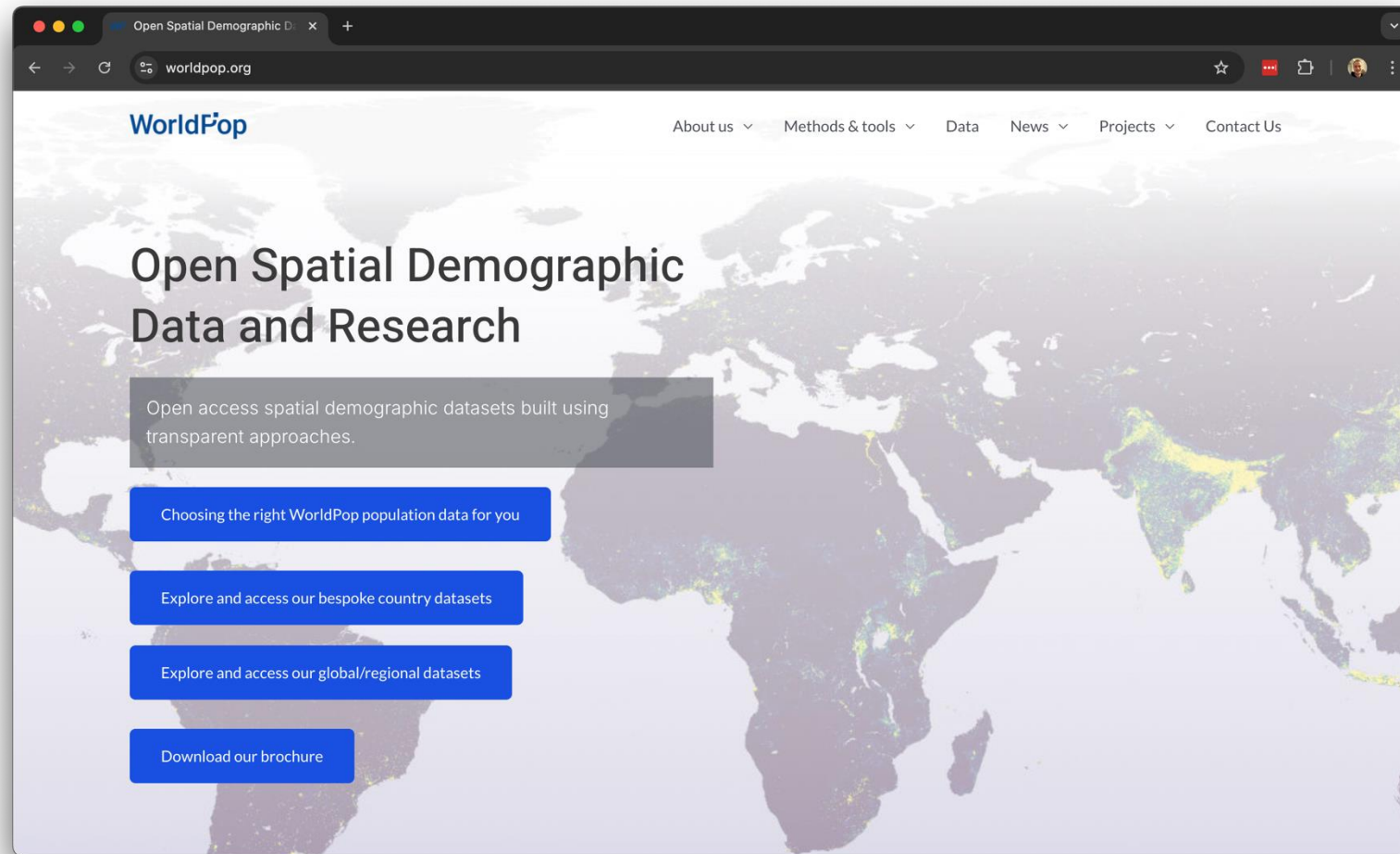






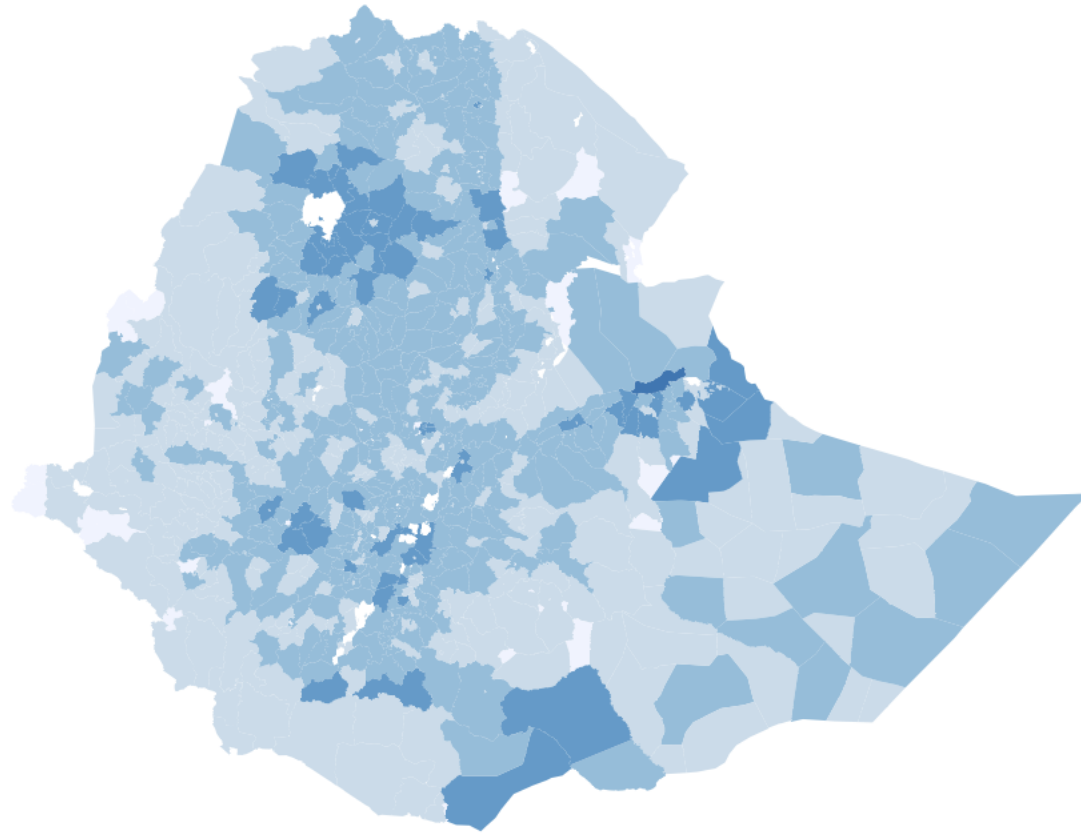


# Dasymetric mapping

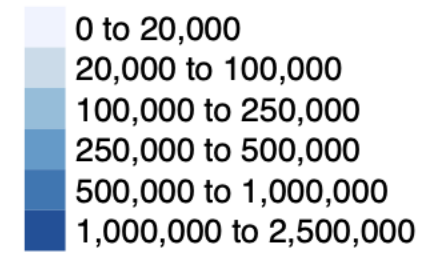




# Raster data

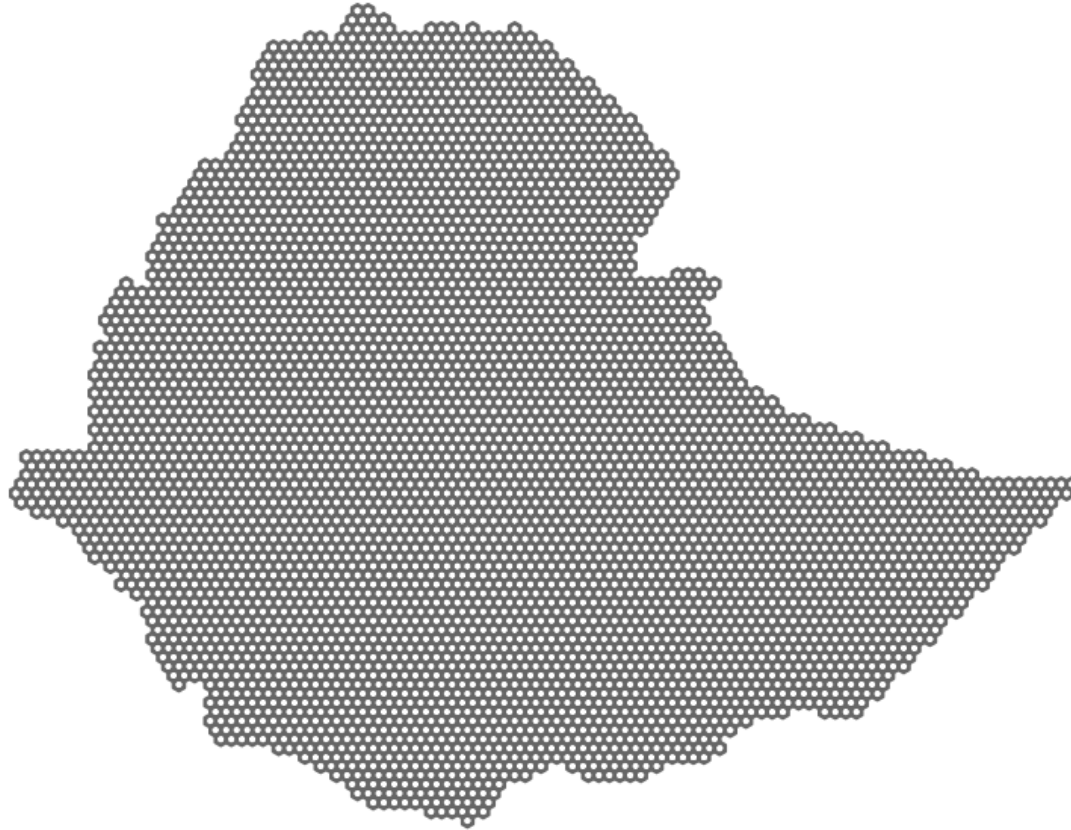


## Population count

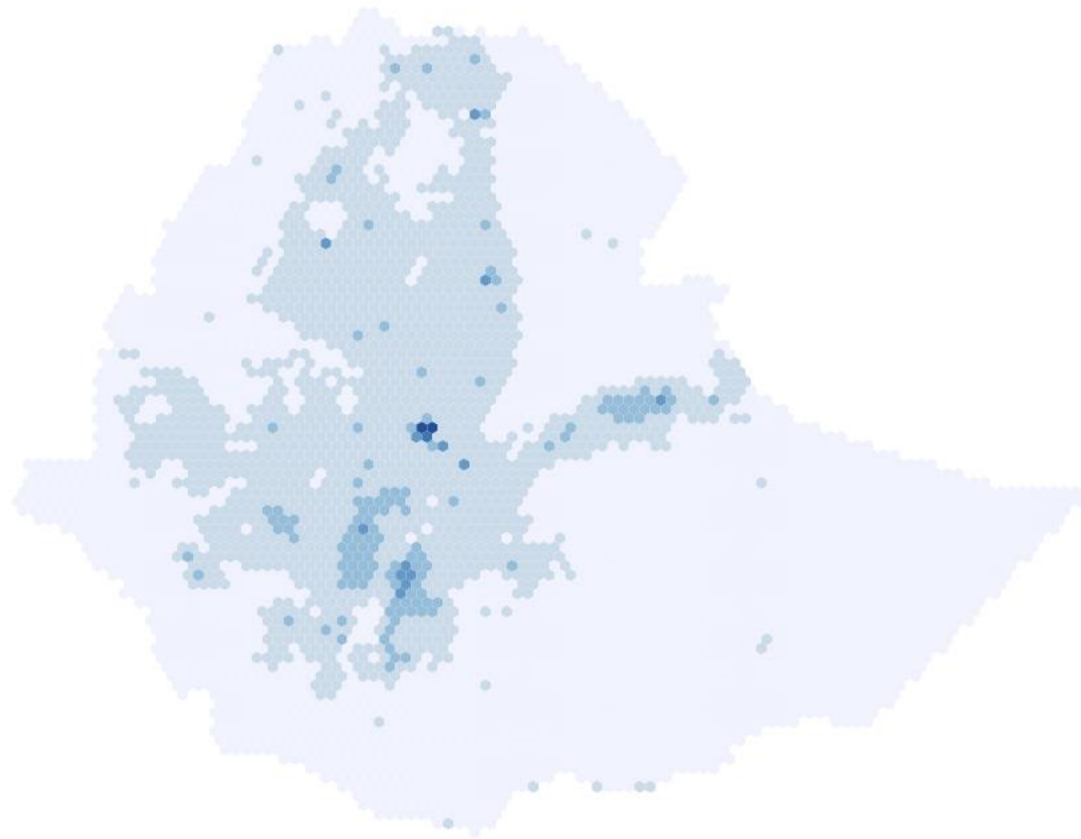




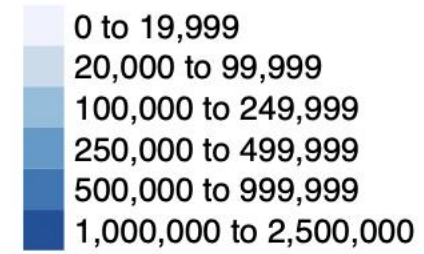
# Raster data



# Raster data



## Population count



# Questions

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# Practical

[jtvandijk.github.io/SA-TIED](https://jtvandijk.github.io/SA-TIED)

