

# 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

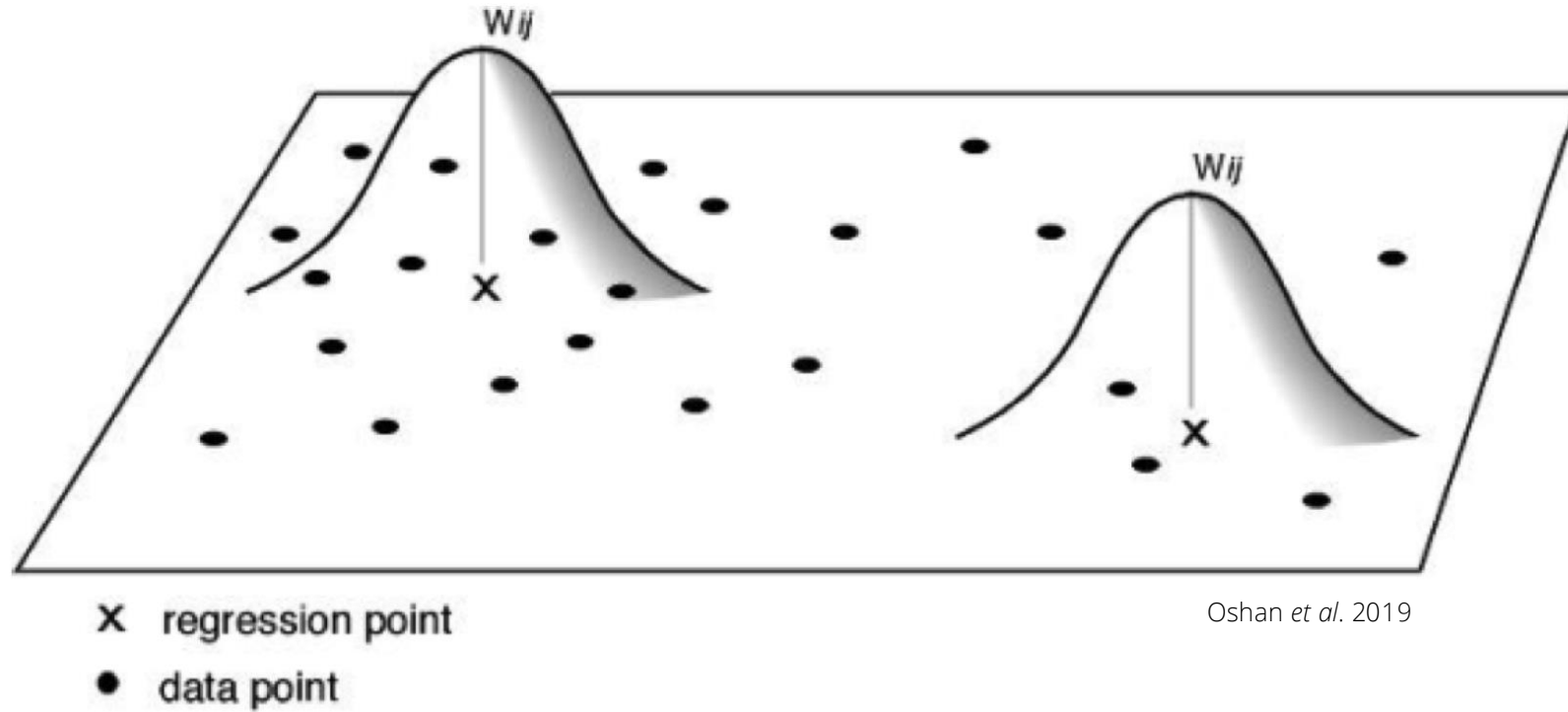
- Unlike traditional global models, which estimate a single set of parameters for the entire study area, geographically weighted statistics allow for parameter estimates that vary across different locations.
- Local means, local standard deviations, local variances.
- Typically uses some **kernel function**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

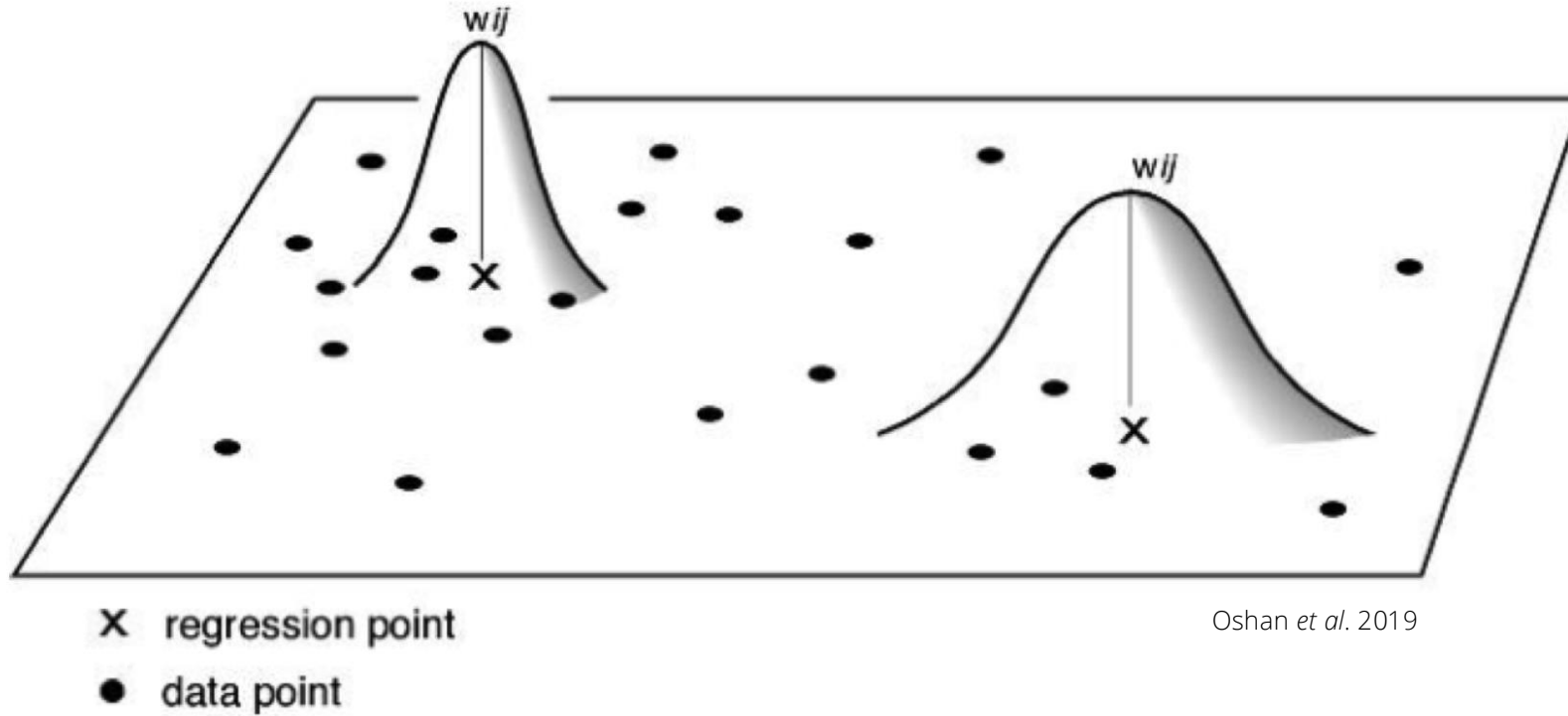
# Geographically weighted statistics



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.

# 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`

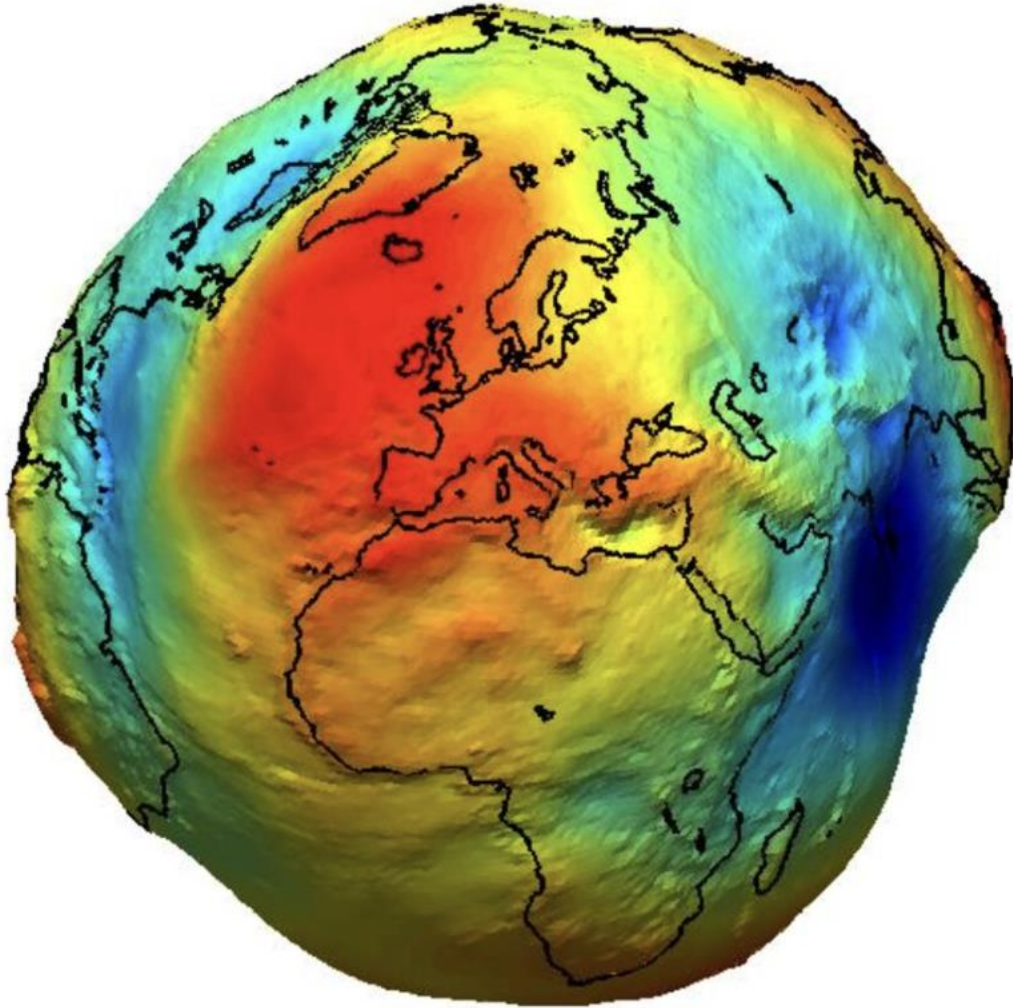
# File formats: Shapefile

- `.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

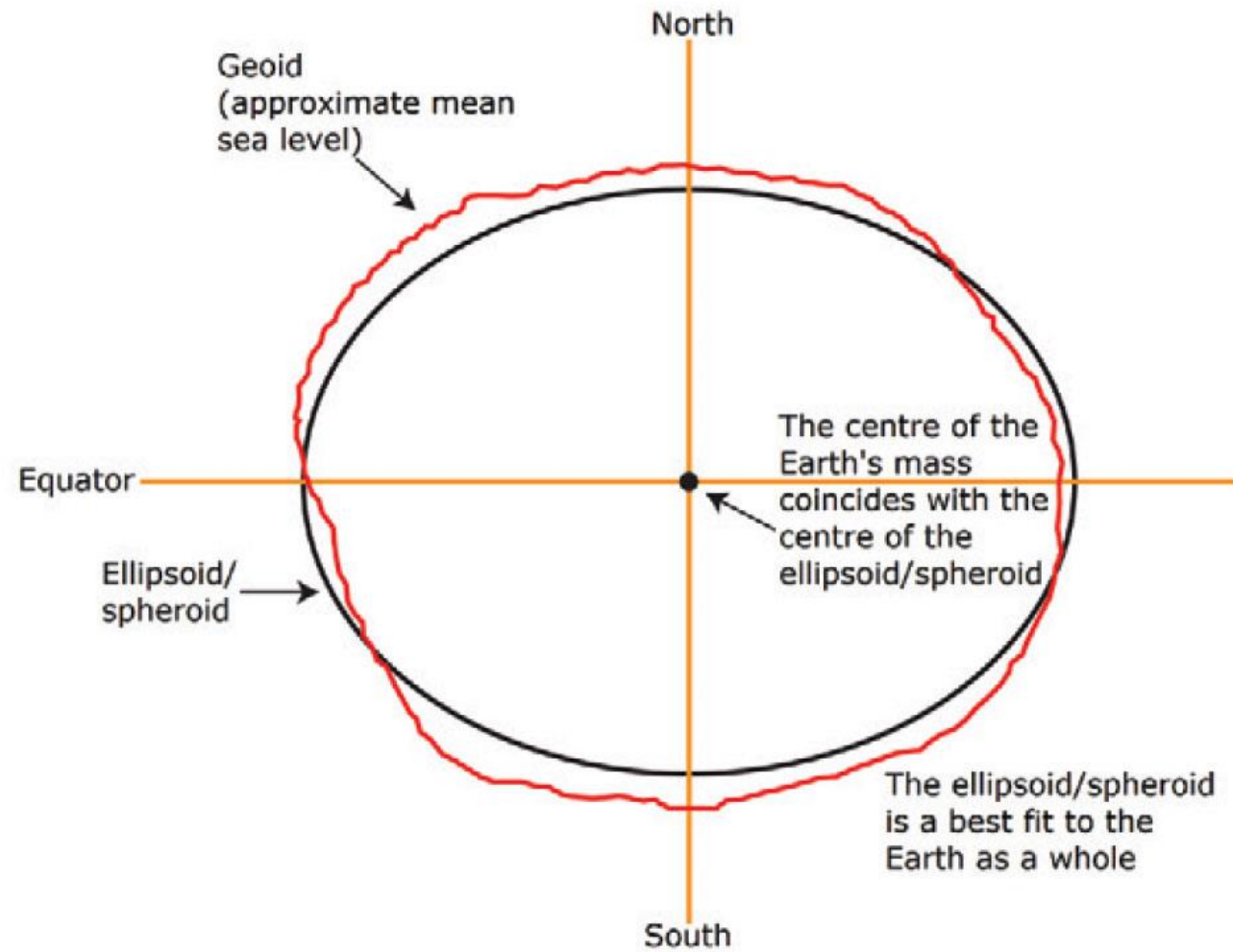


# Map projections

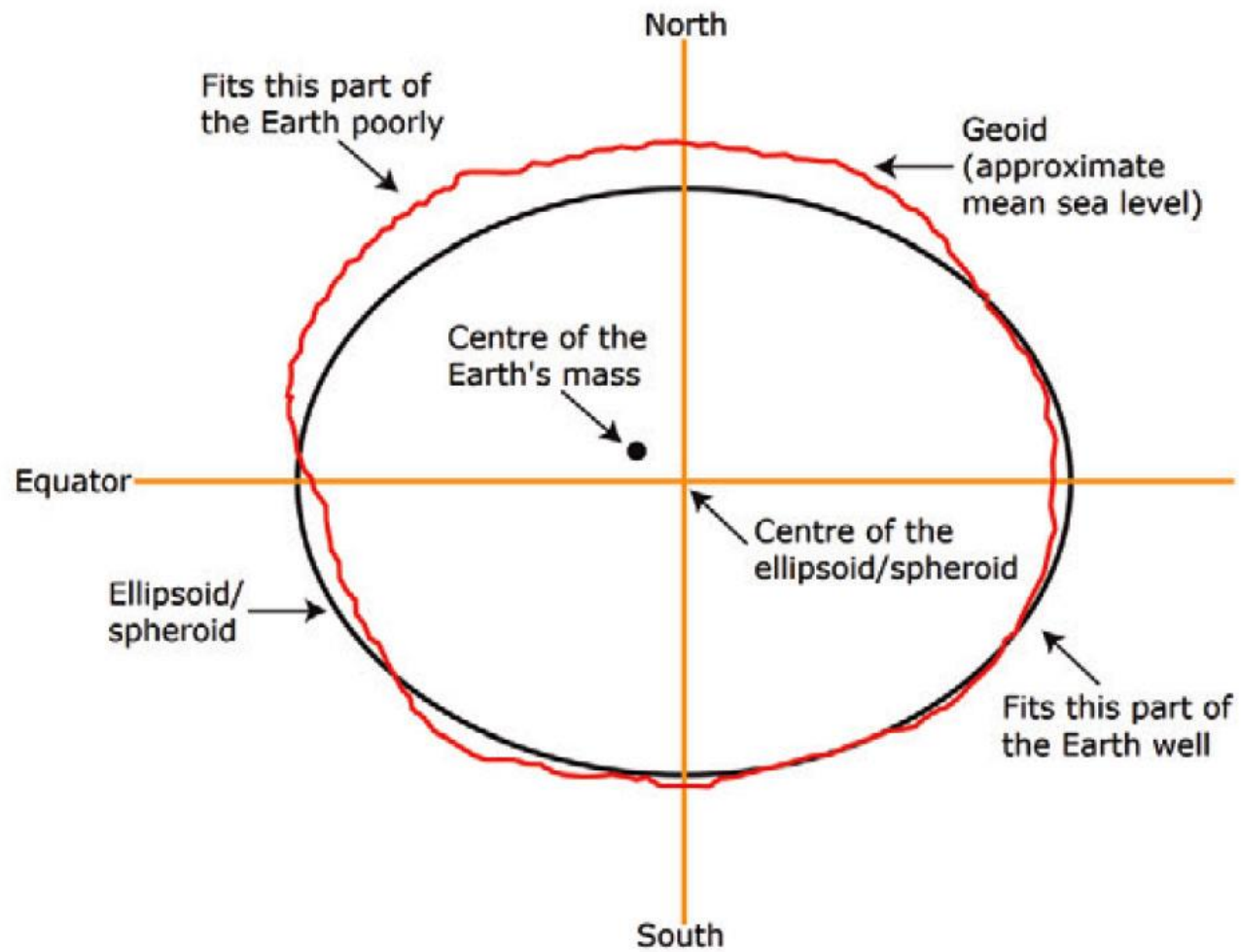




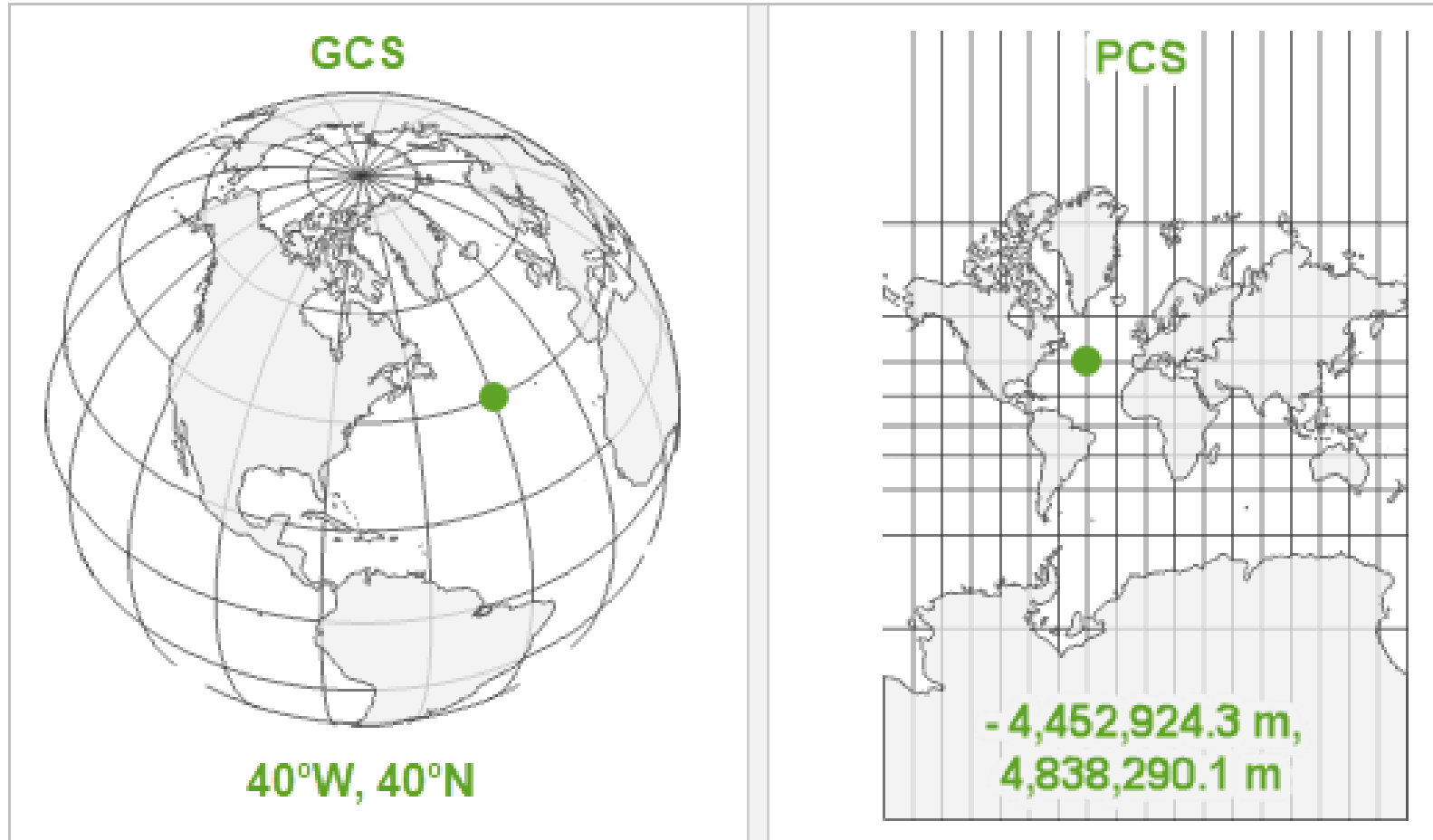
# Map projections



# Map projections



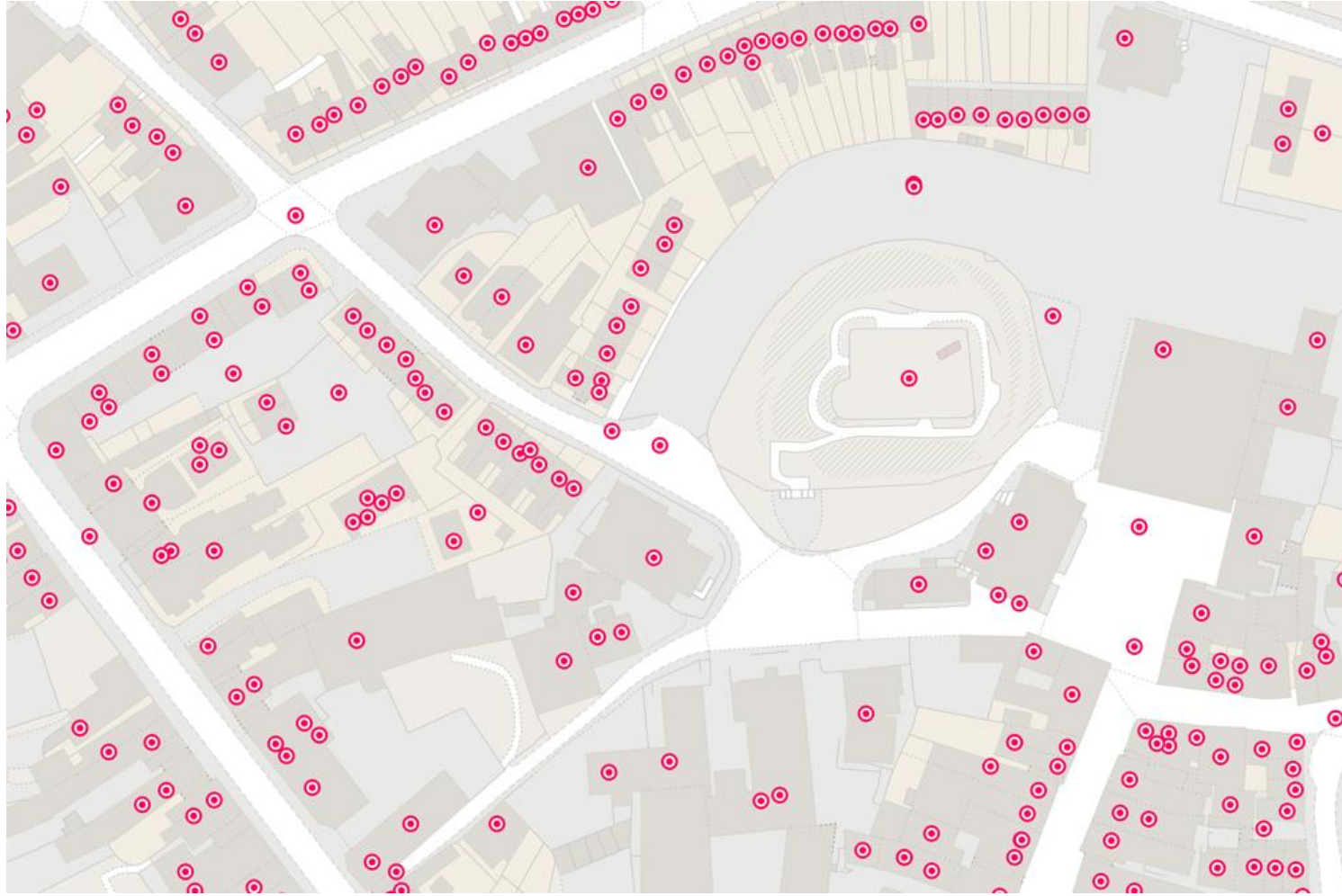
# Map projections



# Geocoding

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

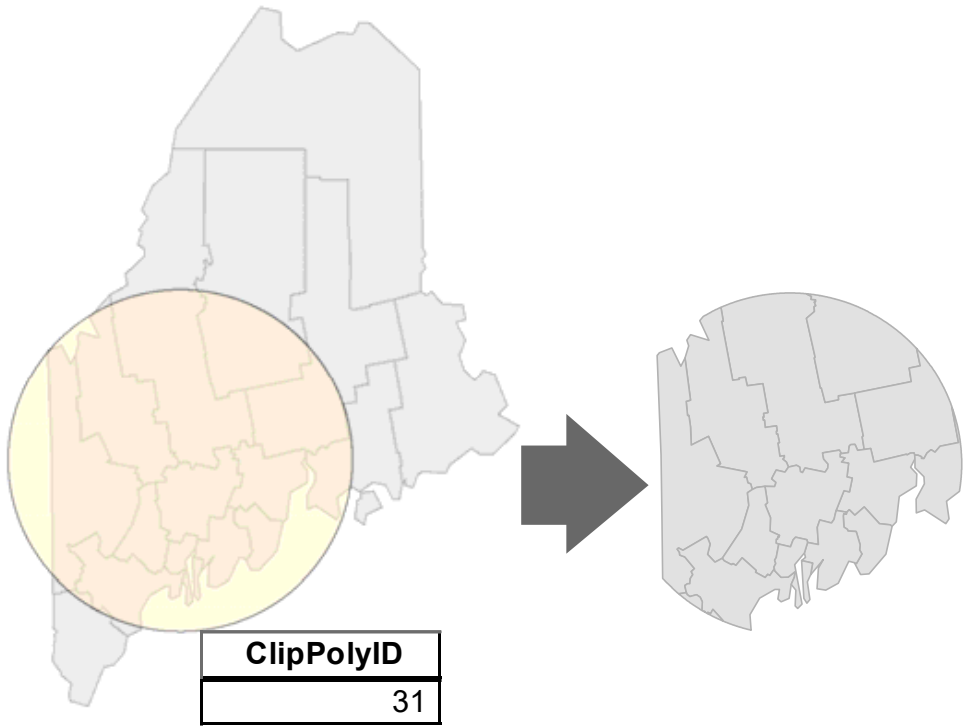
# Geocoding





# Spatial operations

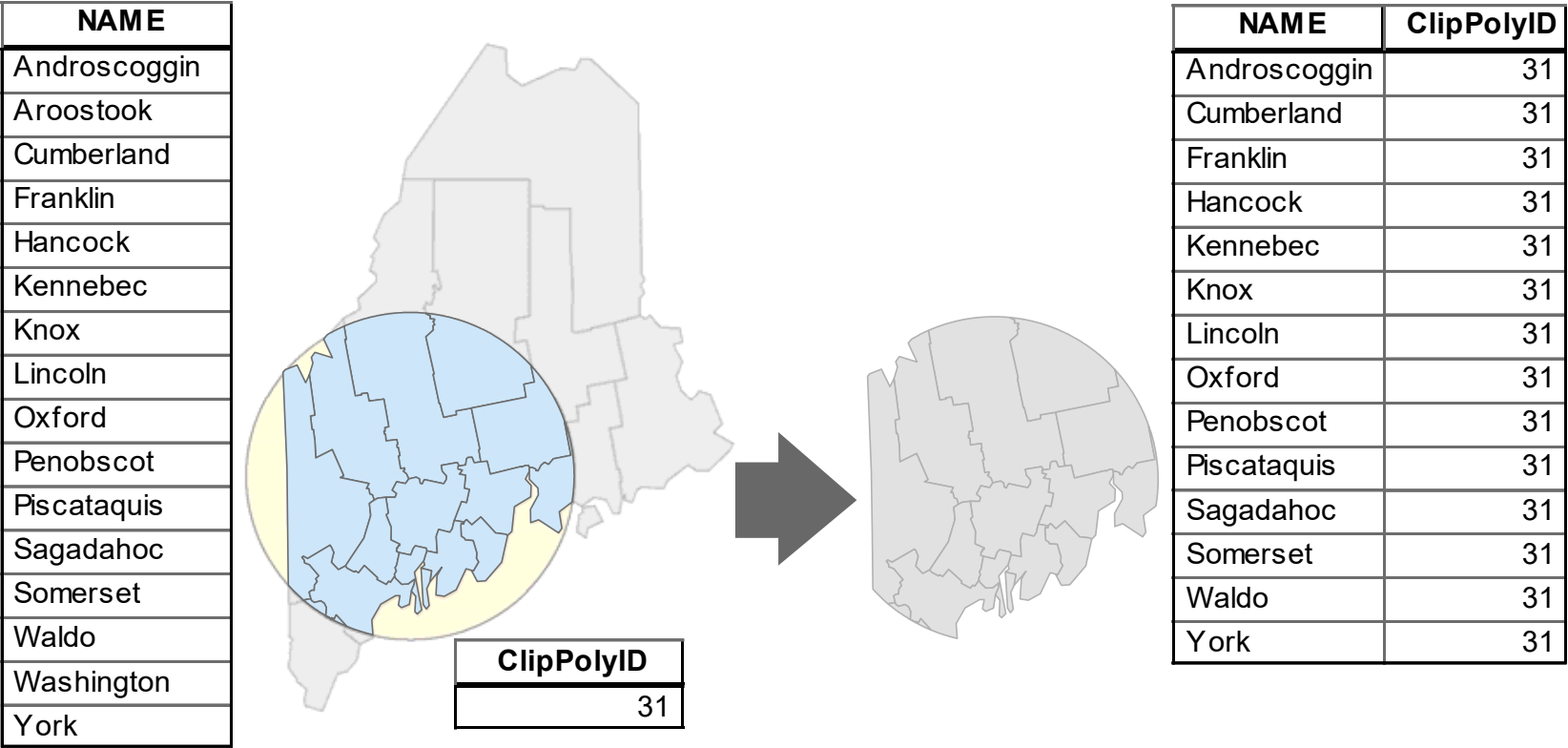
NAME
Androscoggin
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Waldo
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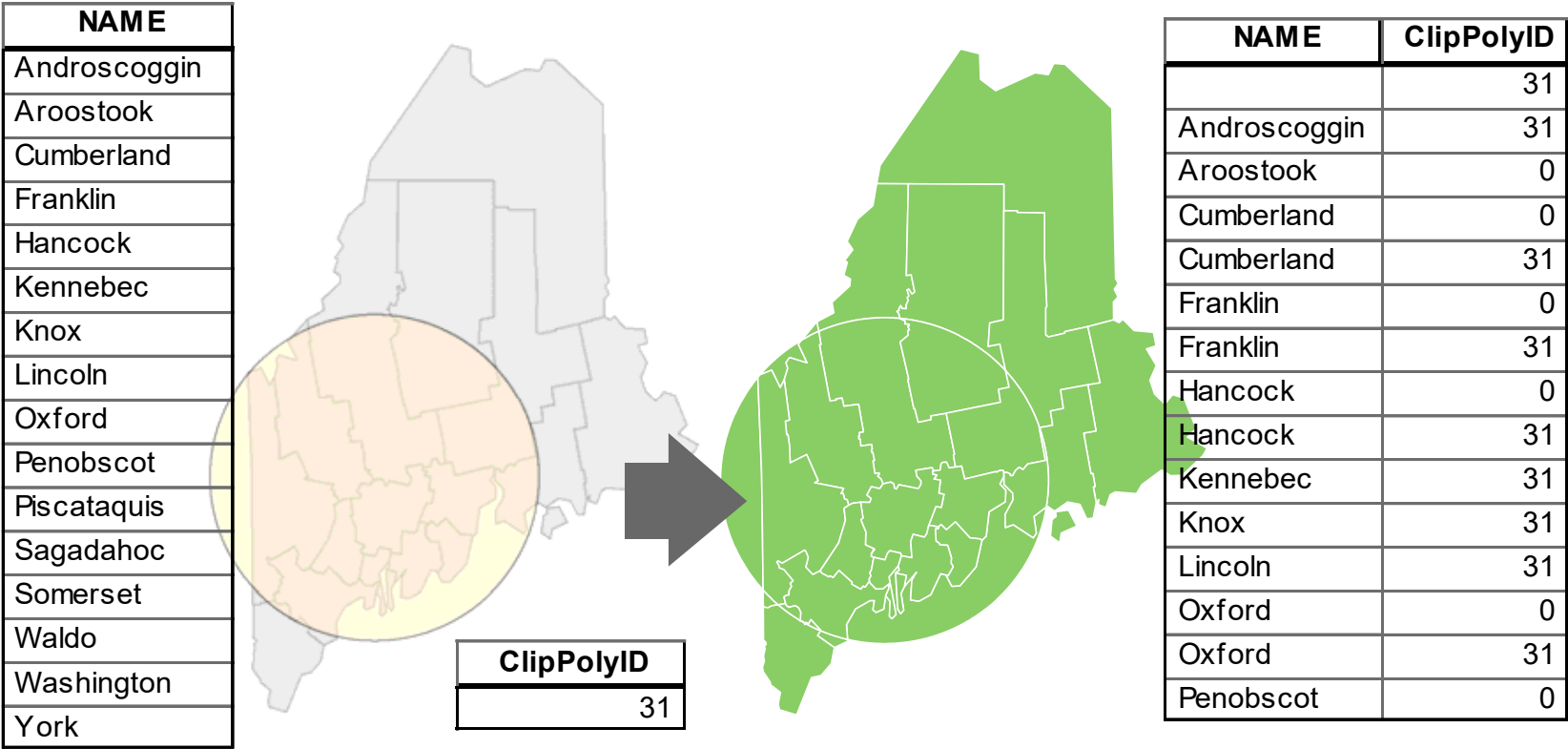
Gimond, M. 2021. Intro to GIS and Spatial Analysis. [online]  
<https://mgimond.github.io/Spatial/introGIS.html>

# Spatial operations



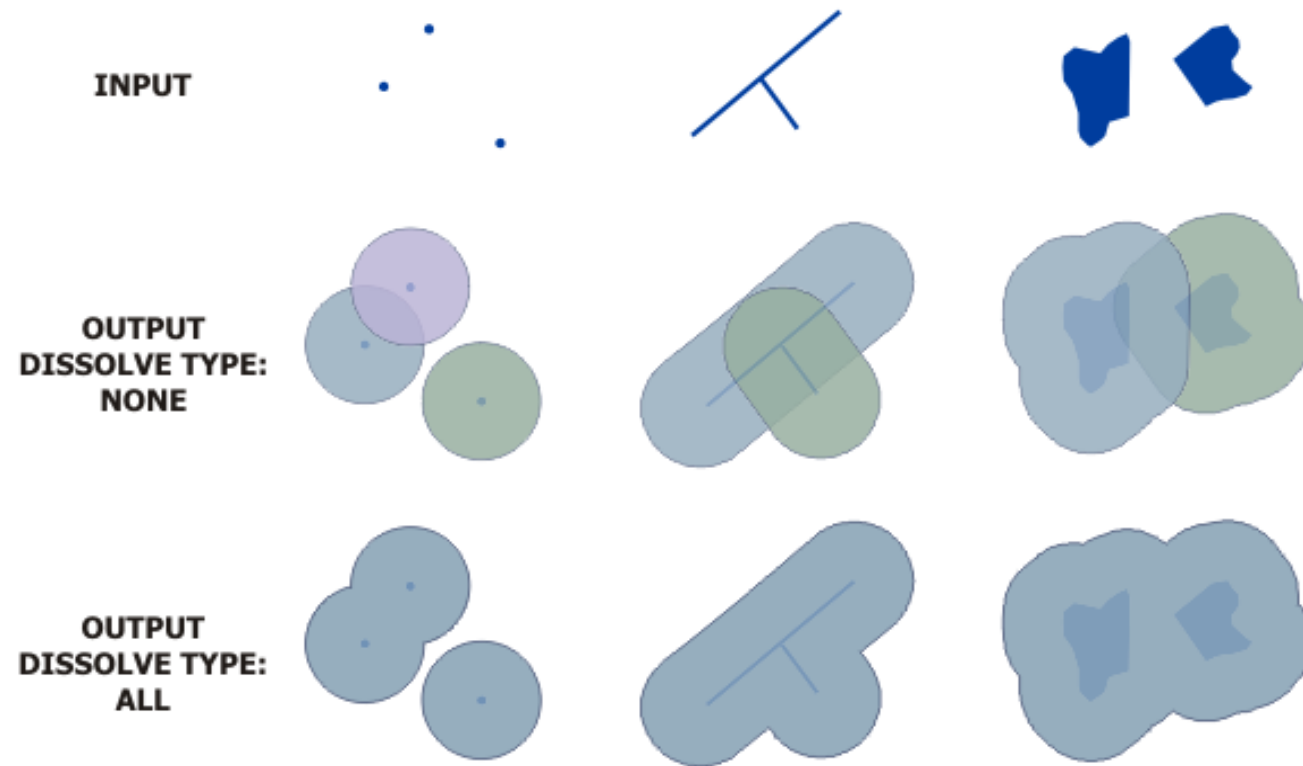
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<https://mgimond.github.io/Spatial/introGIS.html>

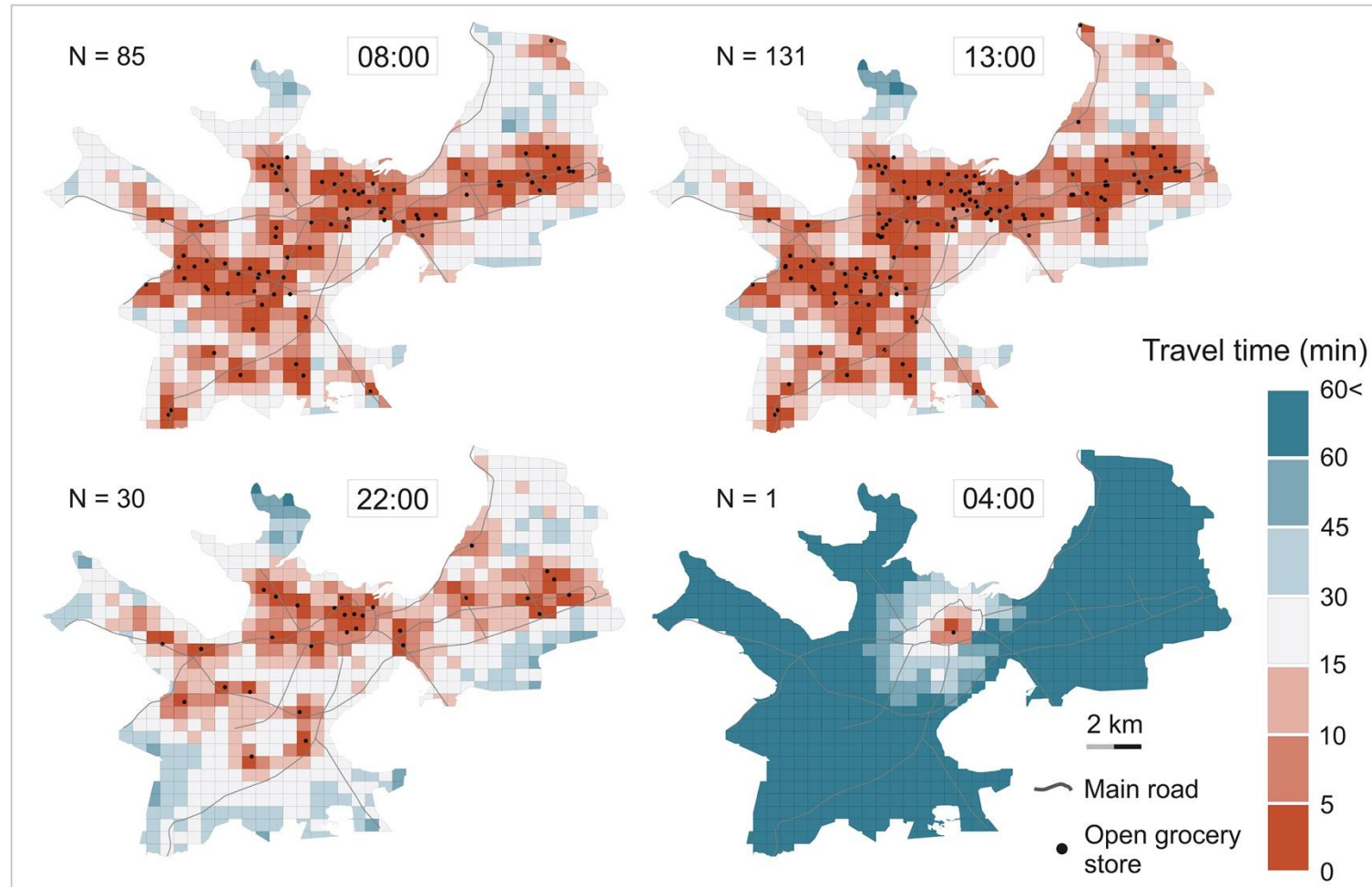
# Spatial operations



# Spatial operations



# Accessibility analysis



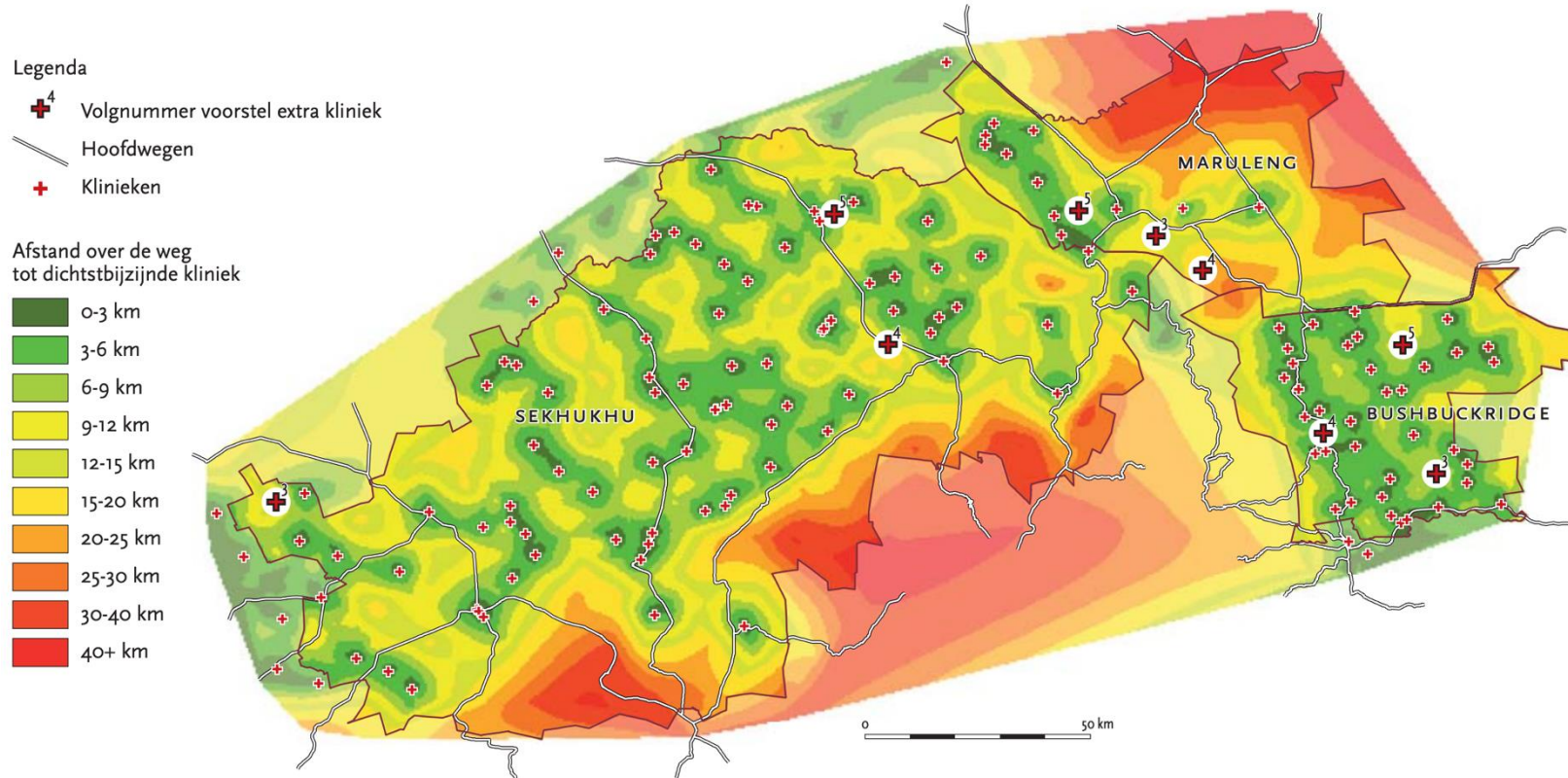
Järv *et al.* 2018

# Accessibility analysis

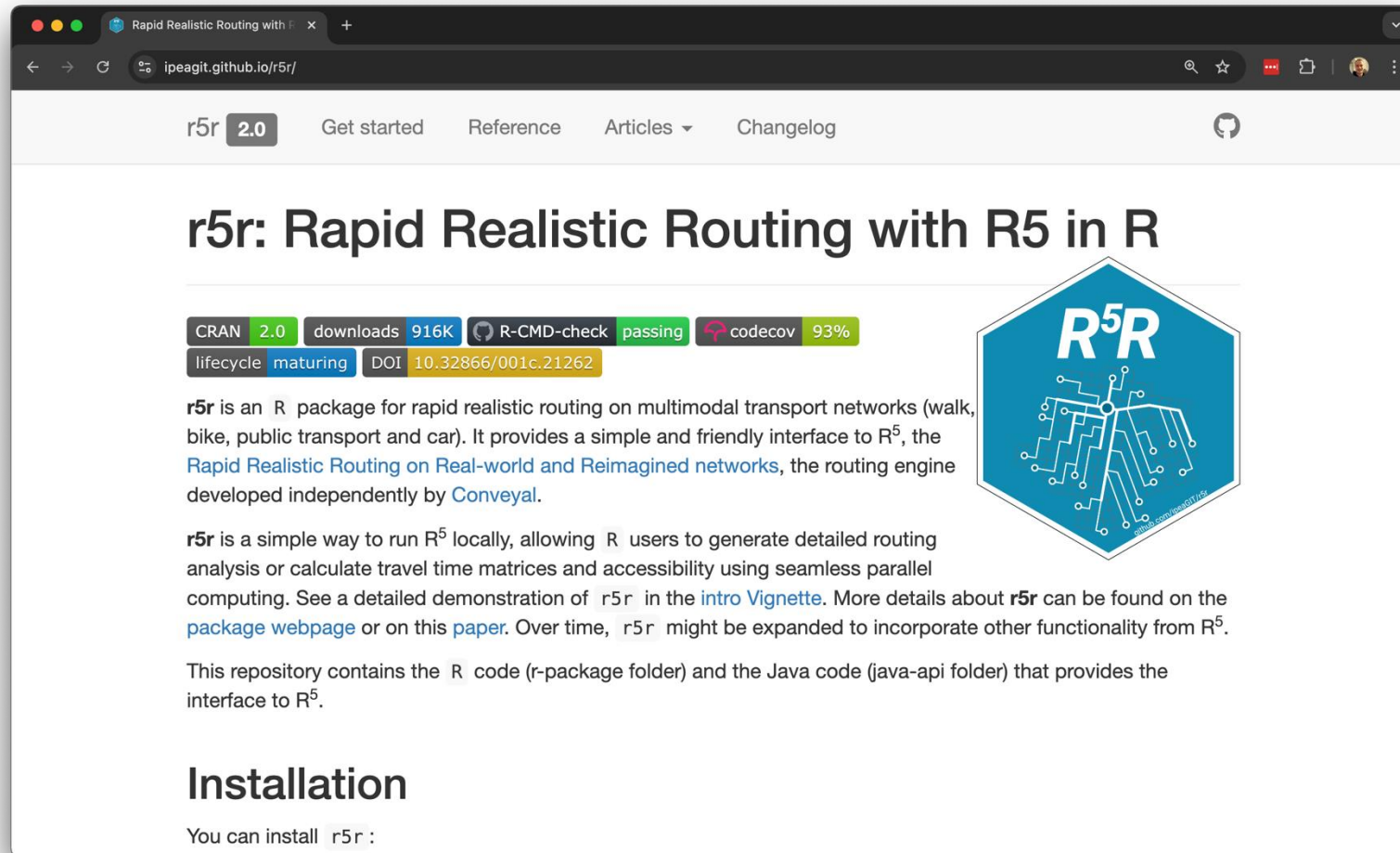




# Accessibility analysis



# Accessibility analysis



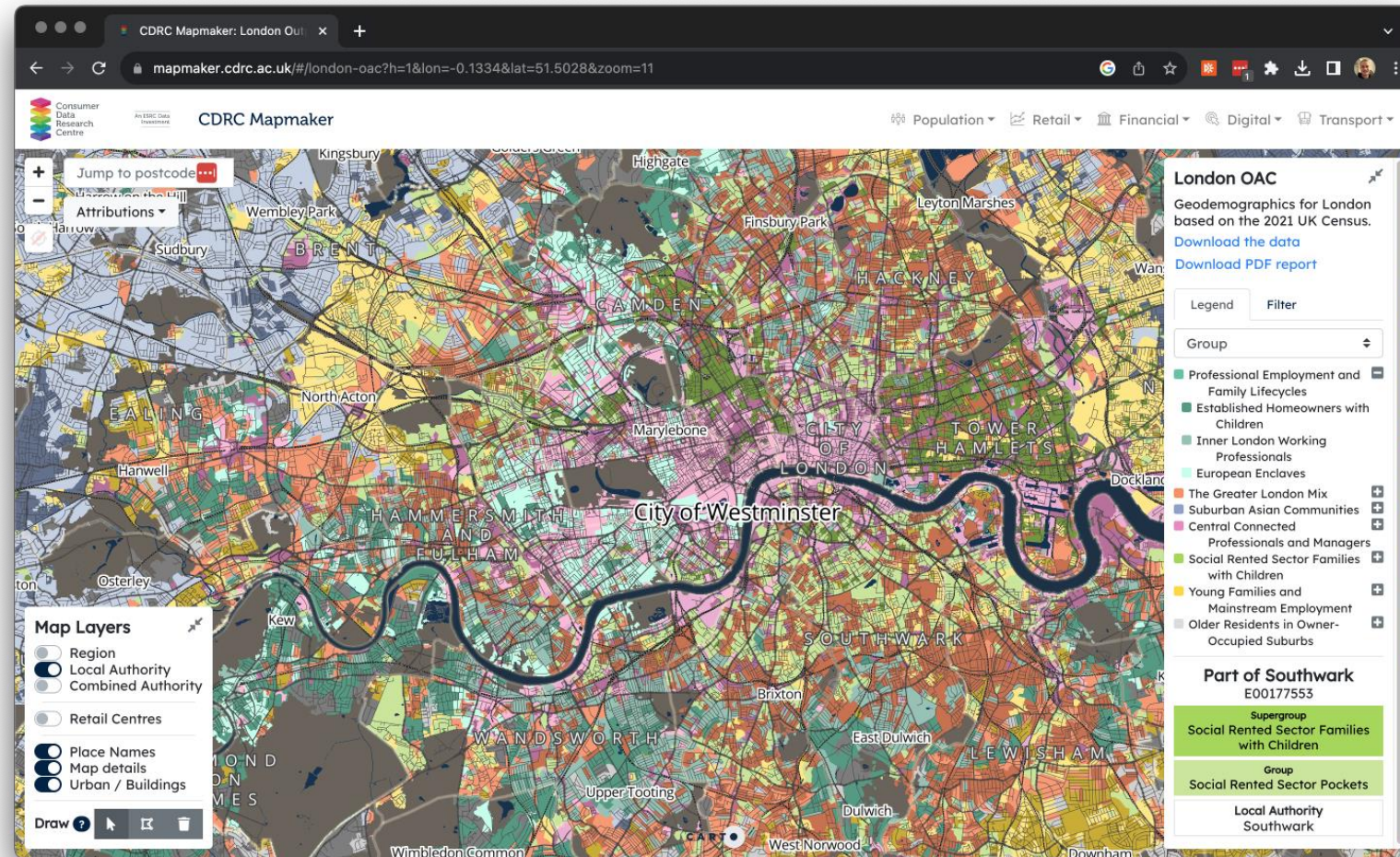
The screenshot shows the GitHub repository page for **r5r** (Rapid Realistic Routing with R5 in R). The page features a navigation bar with links to "Get started", "Reference", "Articles", and "Changelog". The main heading is "r5r: Rapid Realistic Routing with R5 in R". Below the heading, there are several status badges: CRAN 2.0, downloads 916K, R-CMD-check passing, codecov 93%, lifecycle maturing, and DOI 10.32866/001c.21262. To the right of the text is a blue hexagonal logo with the text "R<sup>5</sup>R" and a circuit-like pattern. The text describes **r5r** as an R package for rapid realistic routing on multimodal transport networks, providing a simple interface to R<sup>5</sup>. It mentions that **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. It also states that 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** :



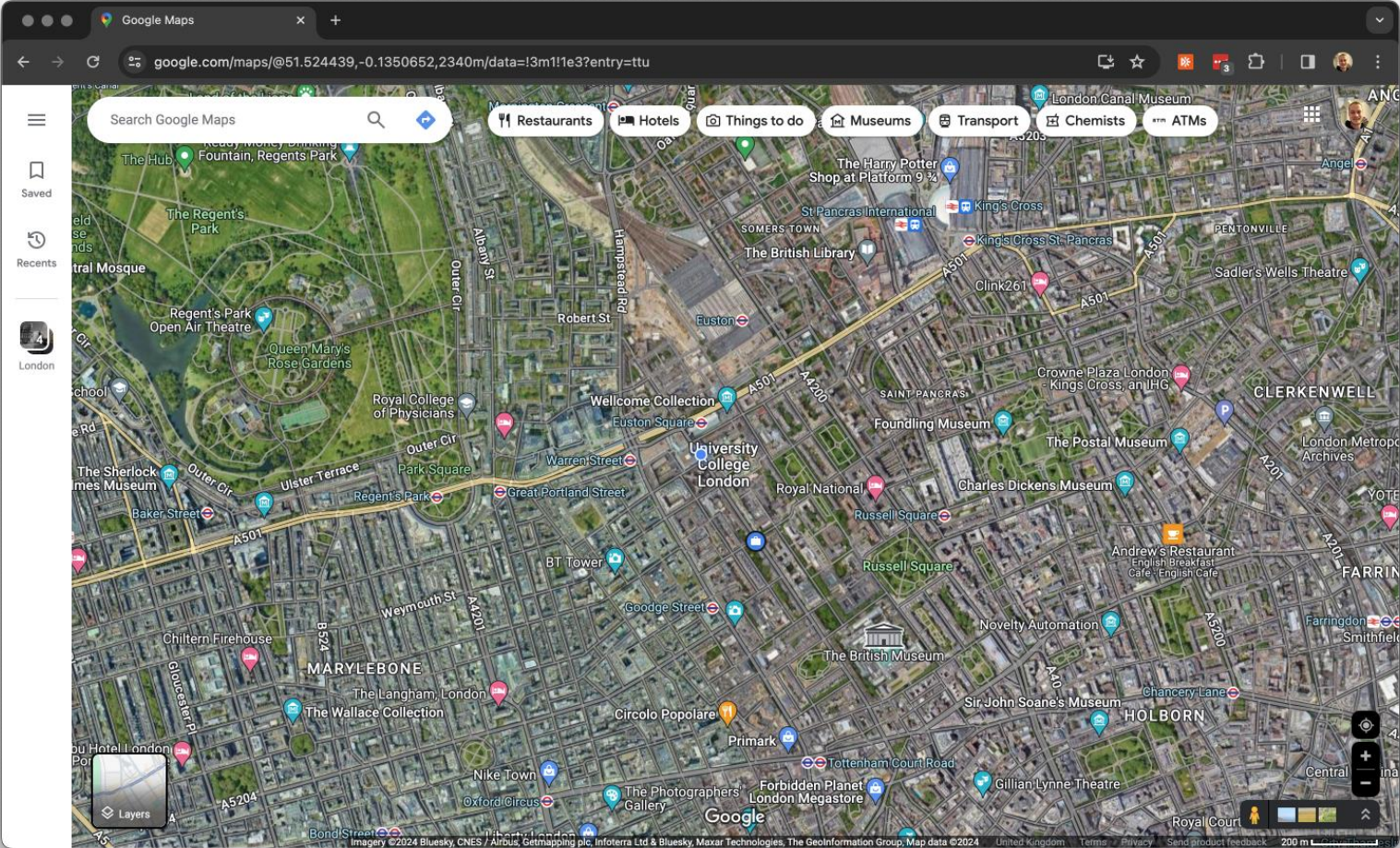
# Geodemographic classification



2021 London Output Area Classification on mapmaker.cdrc.ac.uk

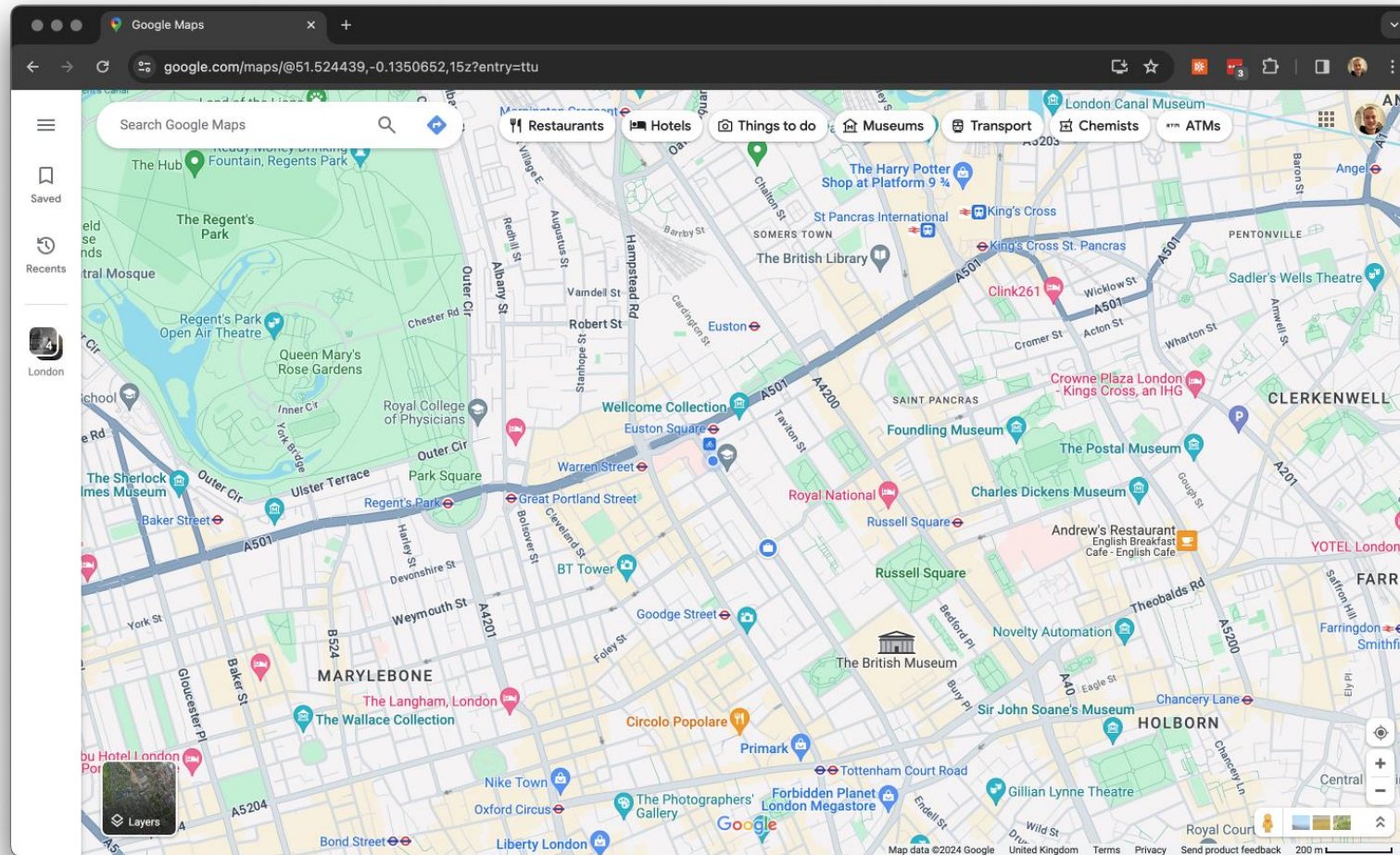


# GIScience





# GI Science



# 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.





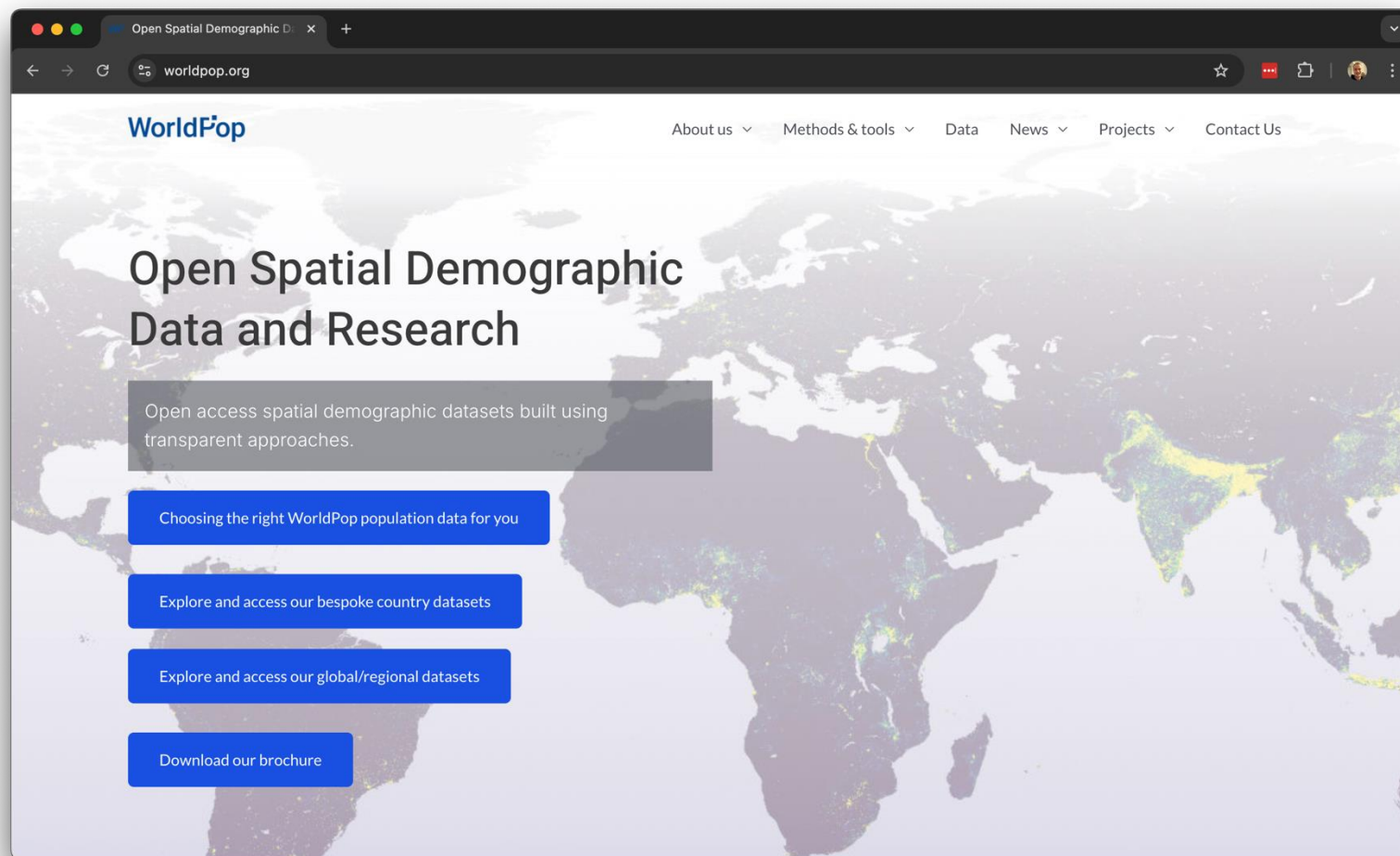






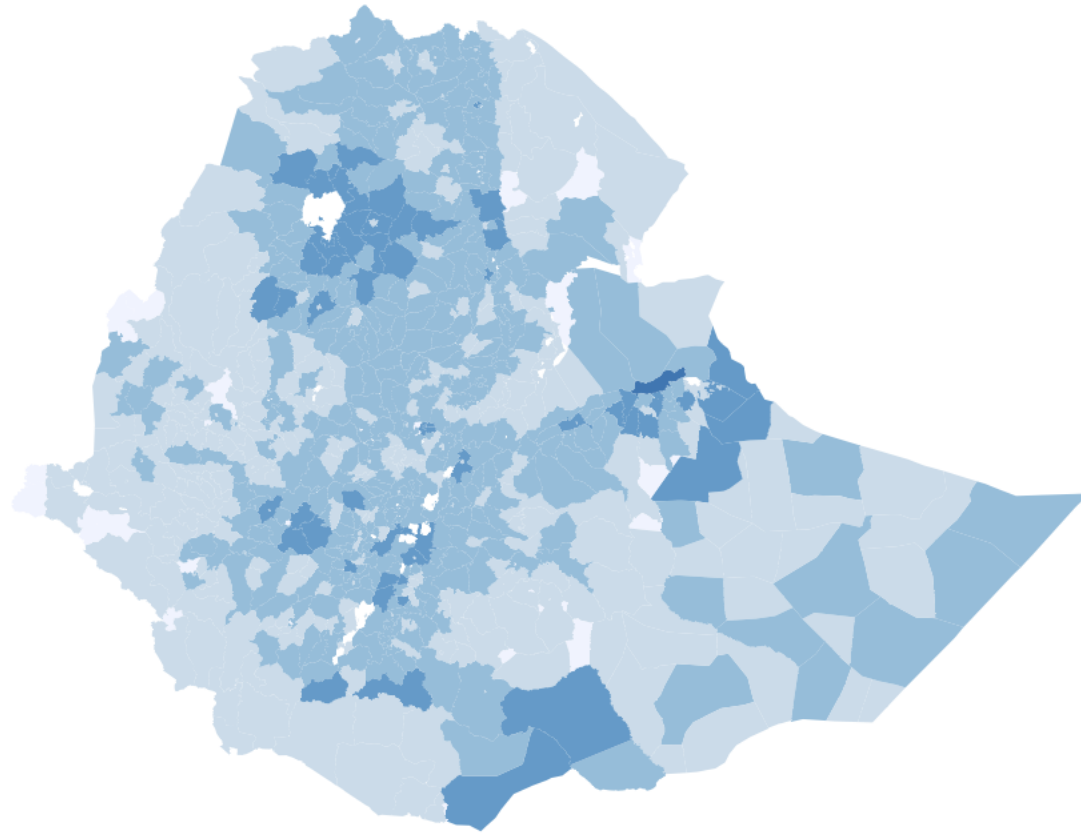


# Dasymetric mapping

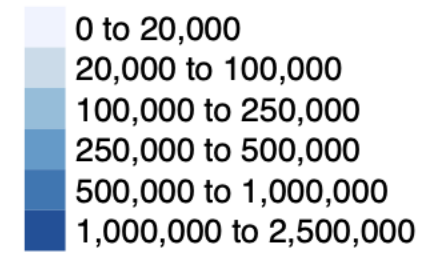




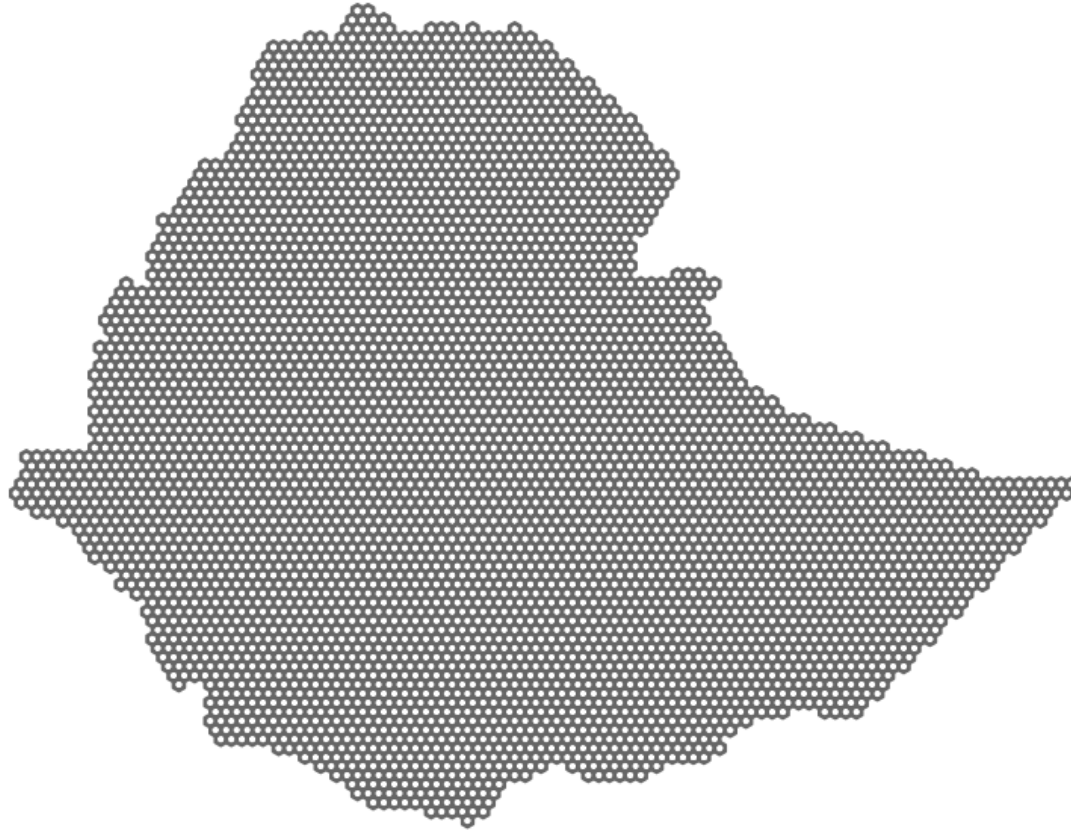
# Raster data



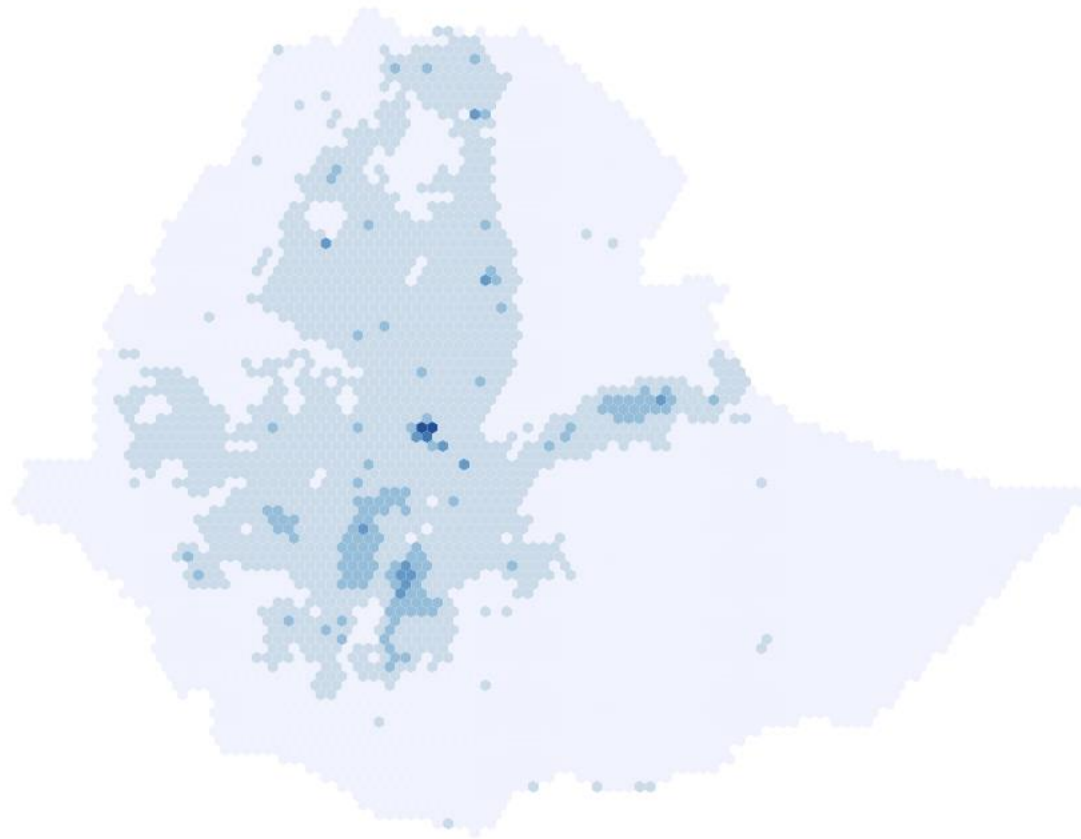
## Population count



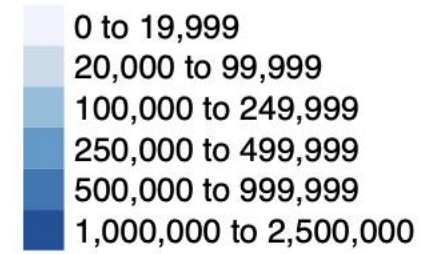
# Raster data



# Raster data



## Population count



# Questions

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