

# Day 1 – Spatial & Spatio-temporal Modelling

## About me

- ▶ Completed a Bachelor's degree in Statistics at FUTA
- ▶ Master's in Mathematical Sciences at AIMS-Tanzania
- ▶ PhD in Statistics and Epidemiology at University of Lancaster
- ▶ Postdoc at University of Manchester
- ▶ Lecturer in Statistics at the University of Manchester



## Overview of the 3 days

- ▶ Spatial and spatio-temporal analysis (different likelihoods)
- ▶ Joint modelling of multiple malaria processes
- ▶ Non-stationary spatial processes
- ▶ Hybrid machine learning + geostatistical models



# Linear Regression

- ▶ Goal: Model a continuous response variable as a linear function of predictors.
- ▶ Model  $Y = X\beta + \epsilon$ , where  $\epsilon \sim N(0, \sigma^2)$
- ▶ Key Assumption
  - ▶ Linearity
  - ▶ Independence
  - ▶ Homoscedasticity (constant variance)
  - ▶ Normality of errors



# Generalized Linear Models (GLMs)

- ▶ Extension of linear models to handle non-normal response distributions.
- ▶ Three components:
  - ▶ Random component: Distribution from the exponential family (e.g., Binomial, Poisson).
  - ▶ Systematic component: Linear predictor  $\eta = X\beta$ .
  - ▶ Link function: Relates  $\mathbb{E}(Y)$  to  $\eta$ .
- ▶ Examples:
  - ▶ Logistic regression for binary outcomes – logit link function
  - ▶ Poisson regression for count data – log link function





## Why Spatial Statistics?

### ► **Spatial Dependence:**

Observations collected at nearby locations are often more similar than those farther apart.

### ► **Ignoring Spatial Structure:**

- Leads to biased parameter estimates.
- Underestimates uncertainty.
- Misses important spatial patterns.

### ► **Applications:**

- Disease mapping
- Environmental monitoring
- Agricultural field trials

### ► **Goal:**

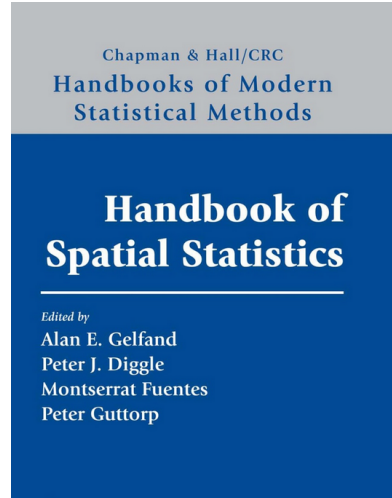
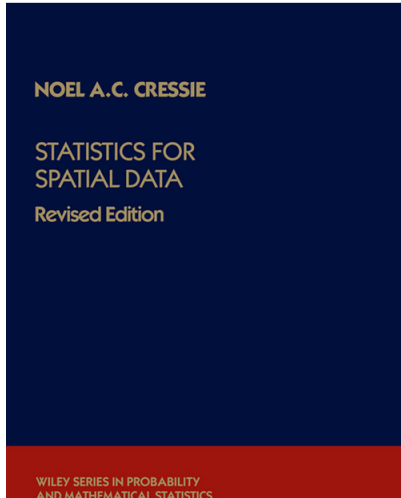
Account for spatial correlation to improve prediction and inference.





# Spatial Analysis

# Spatial Statistics



## Classification of spatial statistics

Cressie's book classifies spatial statistics according to **data format**:

1. Geostatistical data
2. Lattice data
3. Point patterns

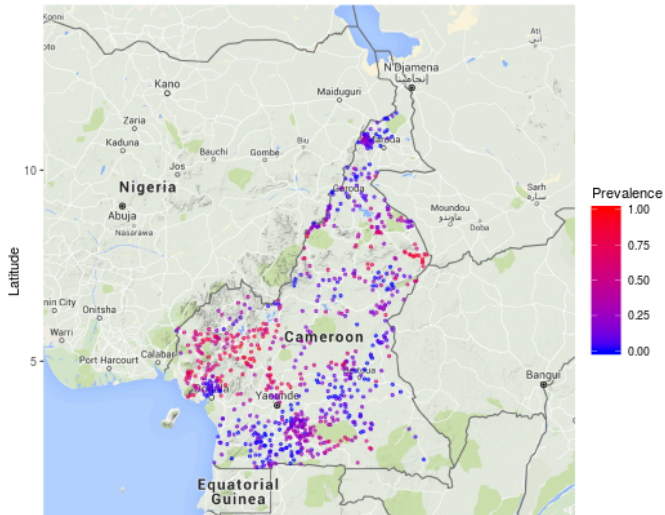
Gelfand's book classifies spatial statistics according to **spatial variation**:

1. Discrete spatial variation
2. Continuous spatial variation

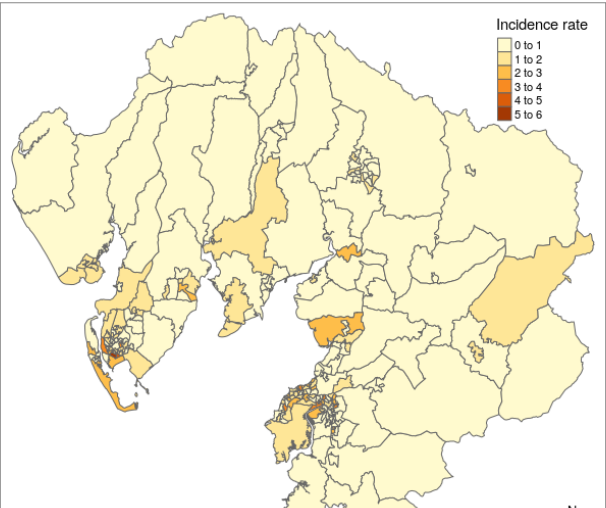




## Geostatistical data: River blindness in Cameroon

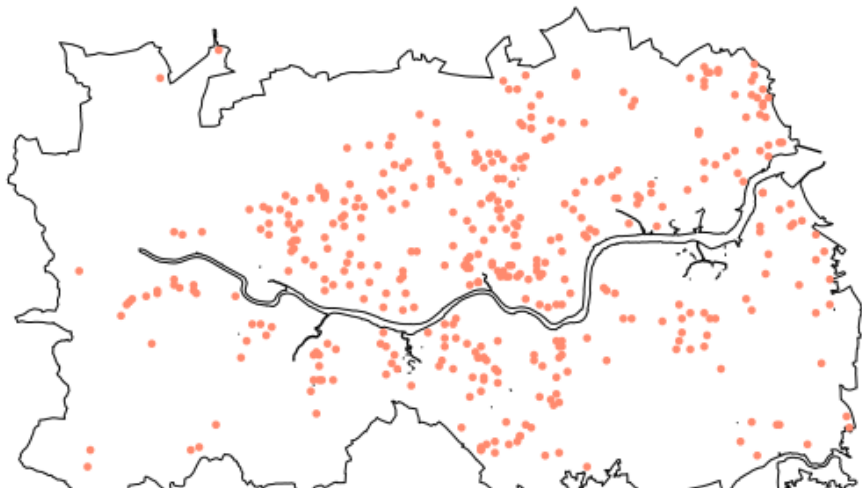


# Lattice Data: COPD emergency admission





## Point pattern: Primary biliary cirrhosis data





# Model-based Geostatistics

## Modelling Geostatistical Data – Model-based Geostatistics

- ▶ The term **Model-based Geostatistics (MBG)** was coined by Peter Diggle in 1998 (Diggle, Tawn, and Moyeed 1998; Diggle and Giorgi 2019).
- ▶ MBG applies general principles of statistical modelling and inference to the analysis of geostatistical data.
- ▶ It emphasises the use of likelihood-based inference.
- ▶ and the use of a latent spatial process (Gaussian or stochastic process)

## References

- Diggle, Peter, and Emanuele Giorgi. 2019. *Model-Based Geostatistics*. CRC Press.
- Diggle, Peter, Jonathan Tawn, and Rana Moyeed. 1998. "Model-Based Geostatistics." *Journal of the Royal Statistical Society: Series C* 47: 299–350.