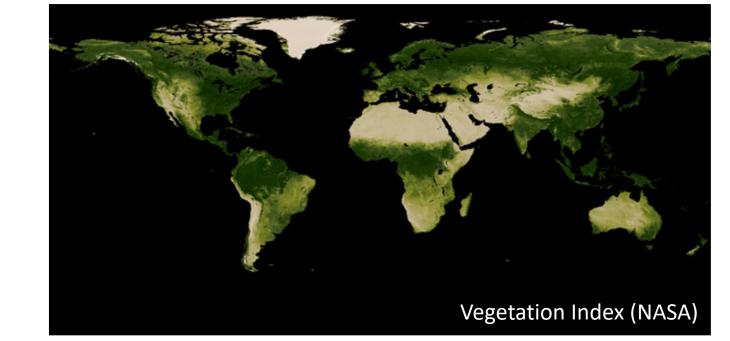


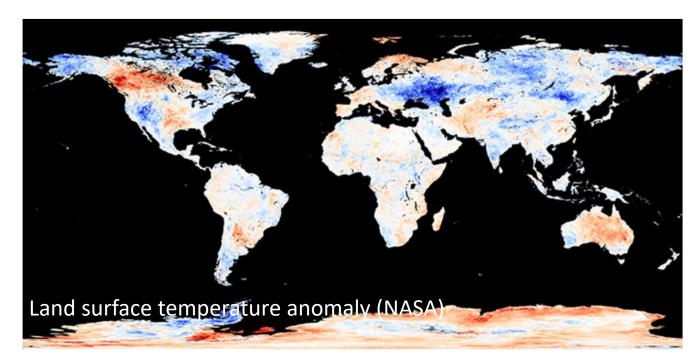
# Comparison of geostatistical methods for spatial data

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## Geospatial data is everywhere

- Climate
- Disease
- Energy
- Satellite data





#### The original problem

- Data at different locations  $(x_i, y_i)$ 
  - $x_i$  location data ( $x_{i1}, x_{i2}$ ), 2D or 3D
  - *y<sub>i</sub>* quantity of interest (dependent variable)
    - Temperature, disease prevalence
- Want to predict  $y_k$  at a different  $x_k$ , where data is not present
- Or map it to a whole region/country
- Covariates of interest
  - Altitude, population density

#### The original problem

- $y_i = f(x_i, u_i, \epsilon_i)$
- $u_i$  covariates
- $\epsilon_i$  error term
- Methods adopt certain assumptions and approximations
- General principle Tobler "everything is related to everything else, but near things are more related than distant things"

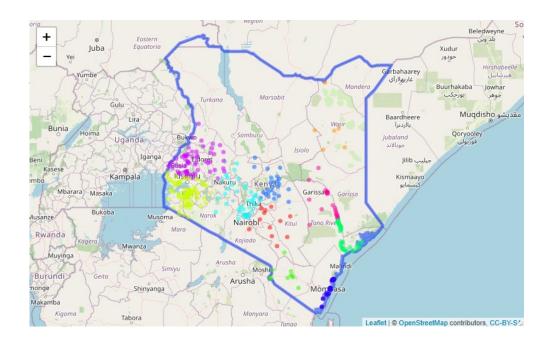
#### Methods we compare

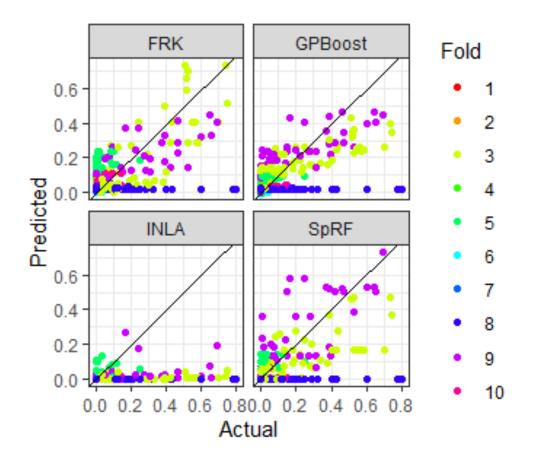
- INLA Integrated Nested Laplace Approximations
- SpRF Spatial Random Forests
- FRK Fixed Rank Kriging
- GPBoost Tree boosting with Gaussian Processes
- Standard parameters for all methods

#### Computation

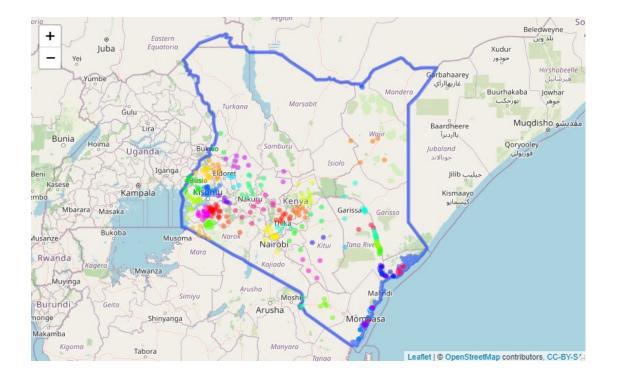
- Data latitude, longitude, number of people tested, number of people positive
- Prevalence = number positive/number tested
- Kenya data in 2009
  - 10 and 50-fold CV
  - Map of Kenya
- Africa
  - Africa maps

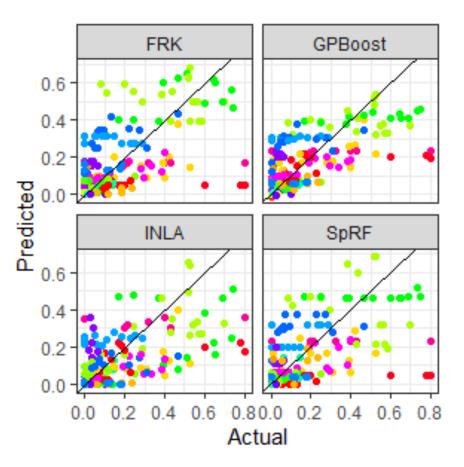
#### 10-fold CV

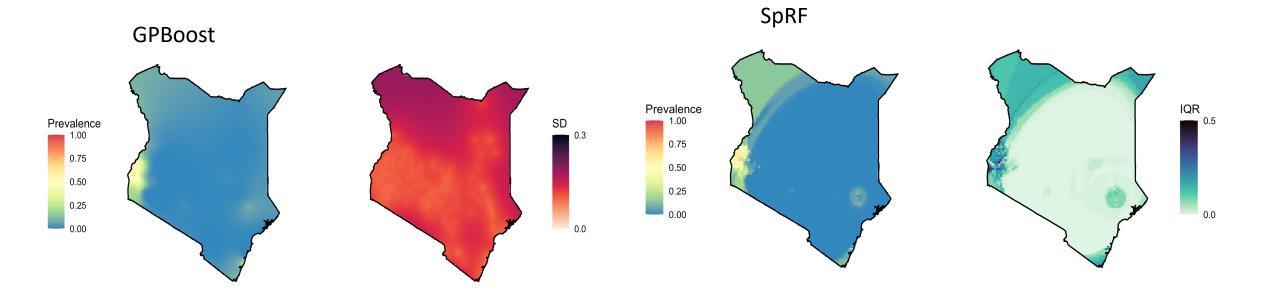




#### 50-fold CV

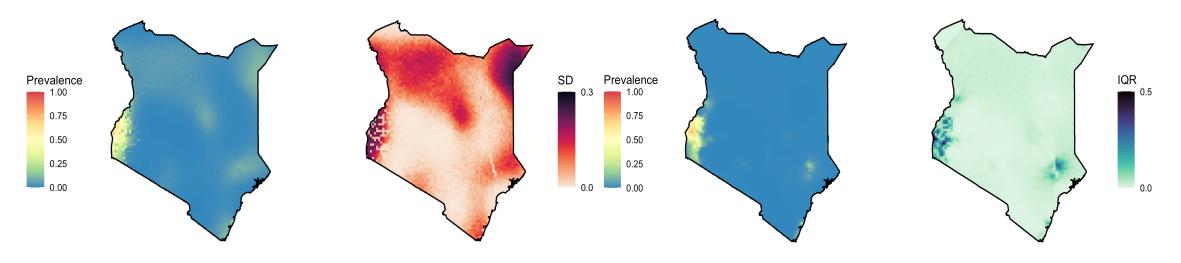


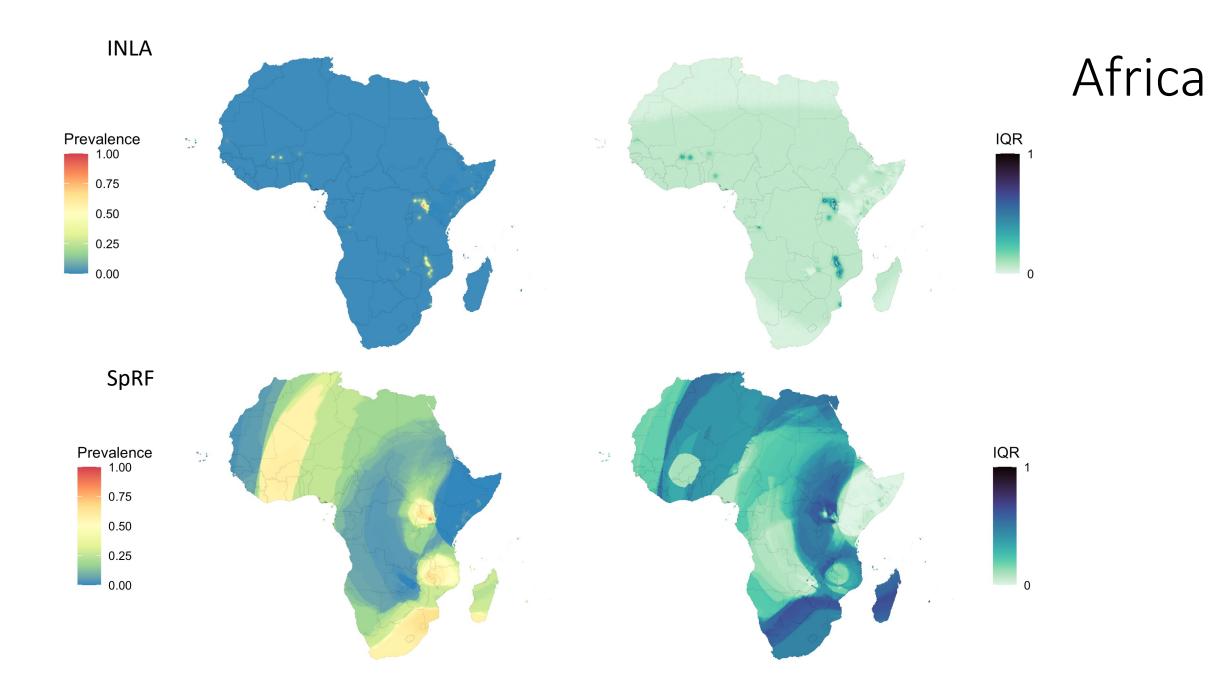


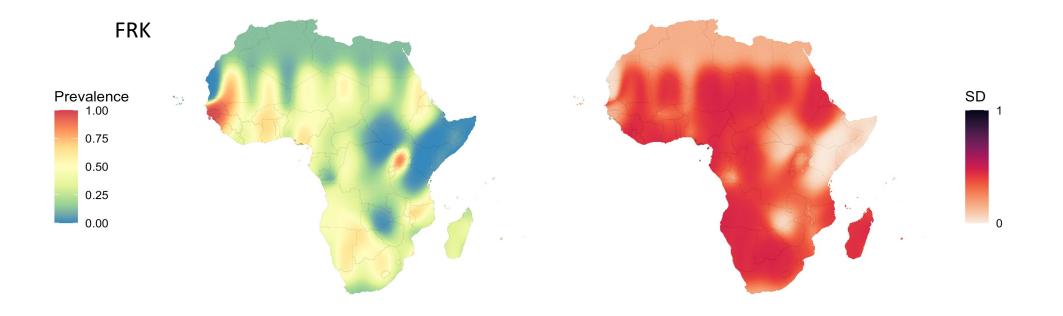


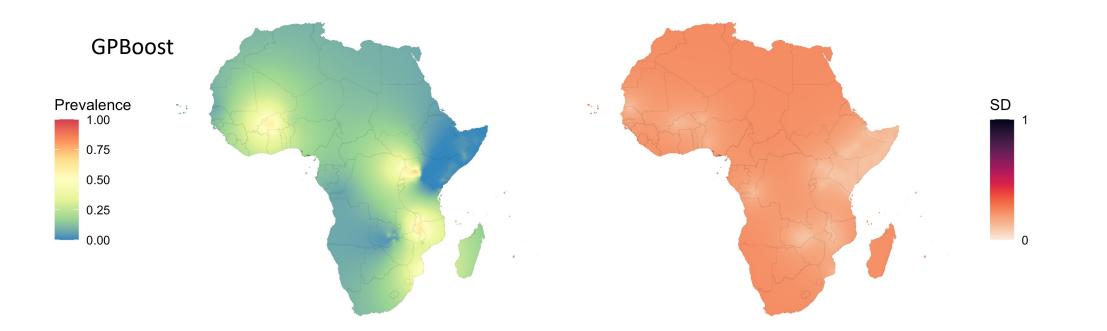
FRK

INLA









#### Reflections

- Maps get a bit messier when we model Africa
- Need to tune the parameters
- Sparsity of points
- See the artefacts of the methods more when you go for bigger scales
- For bigger scale maps users need to understand the methods more
- Not off the shelf— at this point for bigger regions!

### Thank you!

Questions?