

Spatial determinants of Ebola virus disease risk for the West African epidemic

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Workshop on causal adjustment in the presence of spatial dependence
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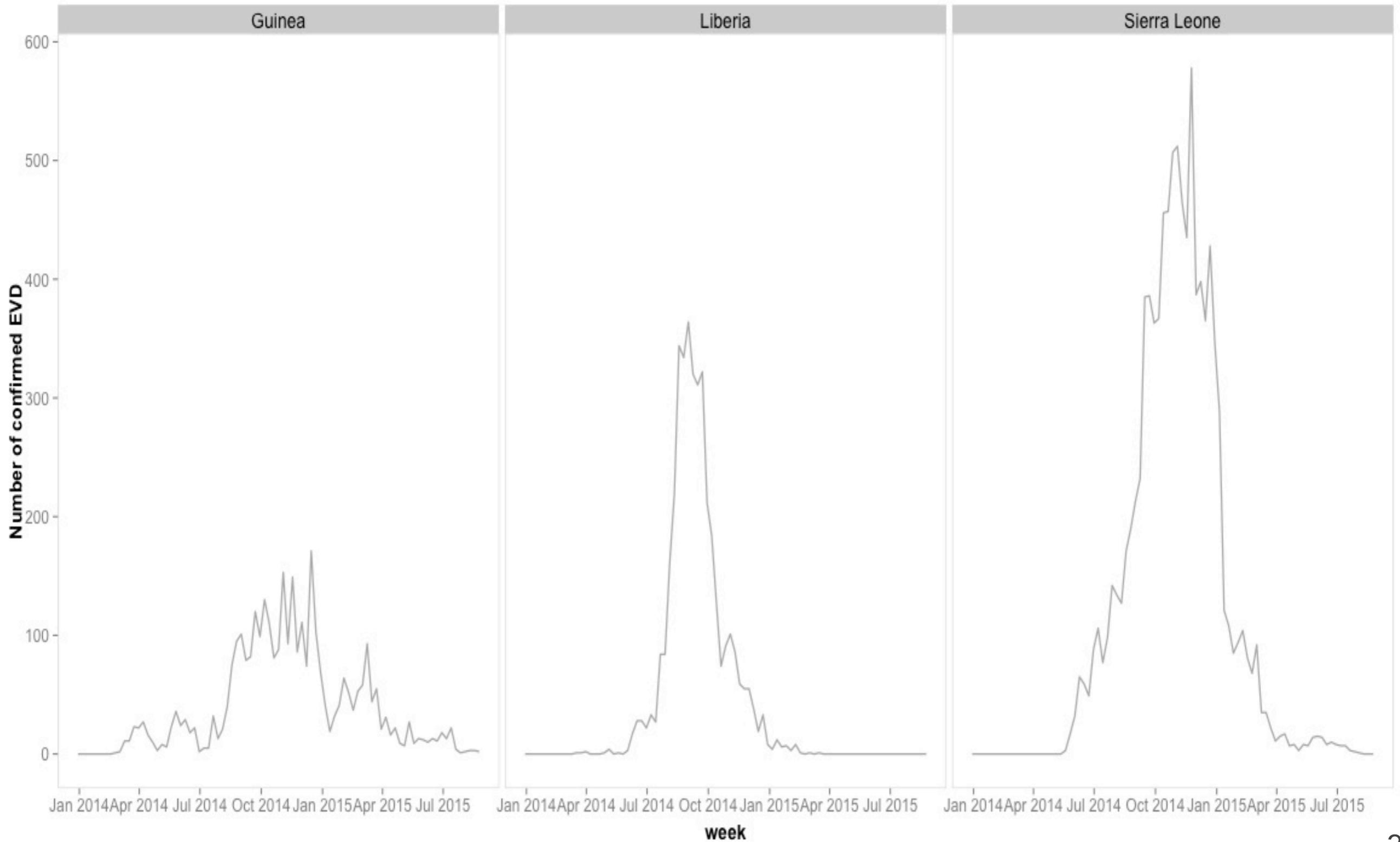
EVD West Africa

- 15,261 confirmed cases with a 74% mortality rate
- Index case: a child who had contact with a fruit bat
- Numerous factors contributed to the human-to-human spread of Ebola virus disease (EVD)
 - Caring for infected individuals, funeral preparations, inadequate healthcare infrastructure
 - Frequency and breadth influenced by population density and mobile populations



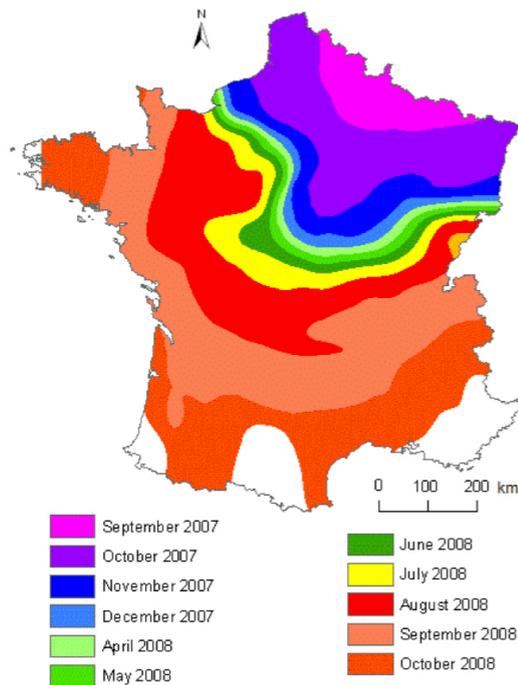
Source: Zoom Dosso/AFP/Getty Images

Epidemic curves 2014-2015



Velocity analysis

- Little is known about the direction and speed of EVD spread
- Velocity is not often calculated for human diseases but has been for several animal diseases
 - Animals: rabies, plague, bluetongue
 - Humans: sleeping sickness, Hanta virus



Source: Pioz et al, Estimating front-wave velocity of infectious diseases: a simple, efficient method applied to bluetongue



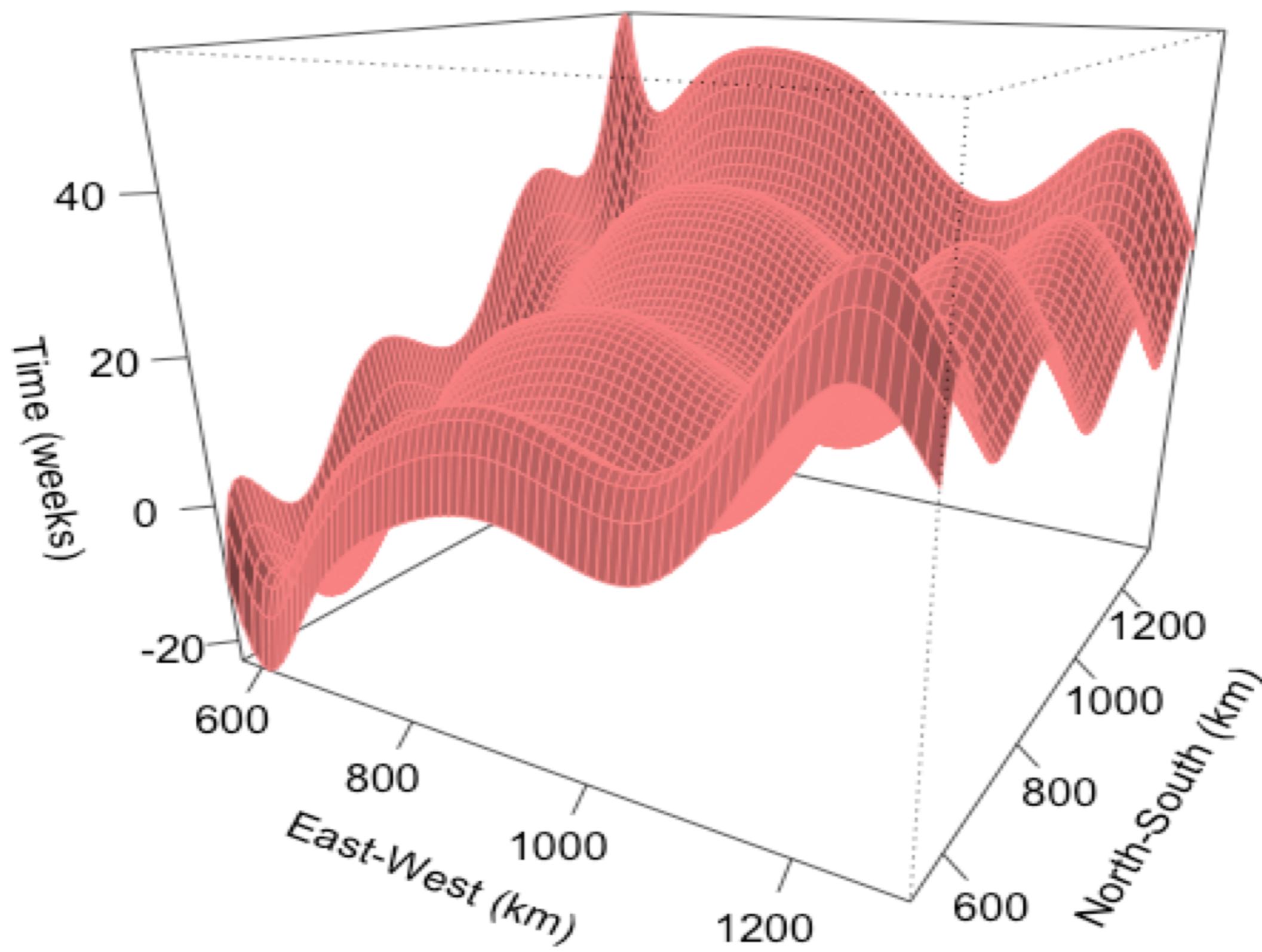
Source: Berrang-Ford L et al, Spatial analysis of sleeping sickness, southeastern Uganda, 1970-2003

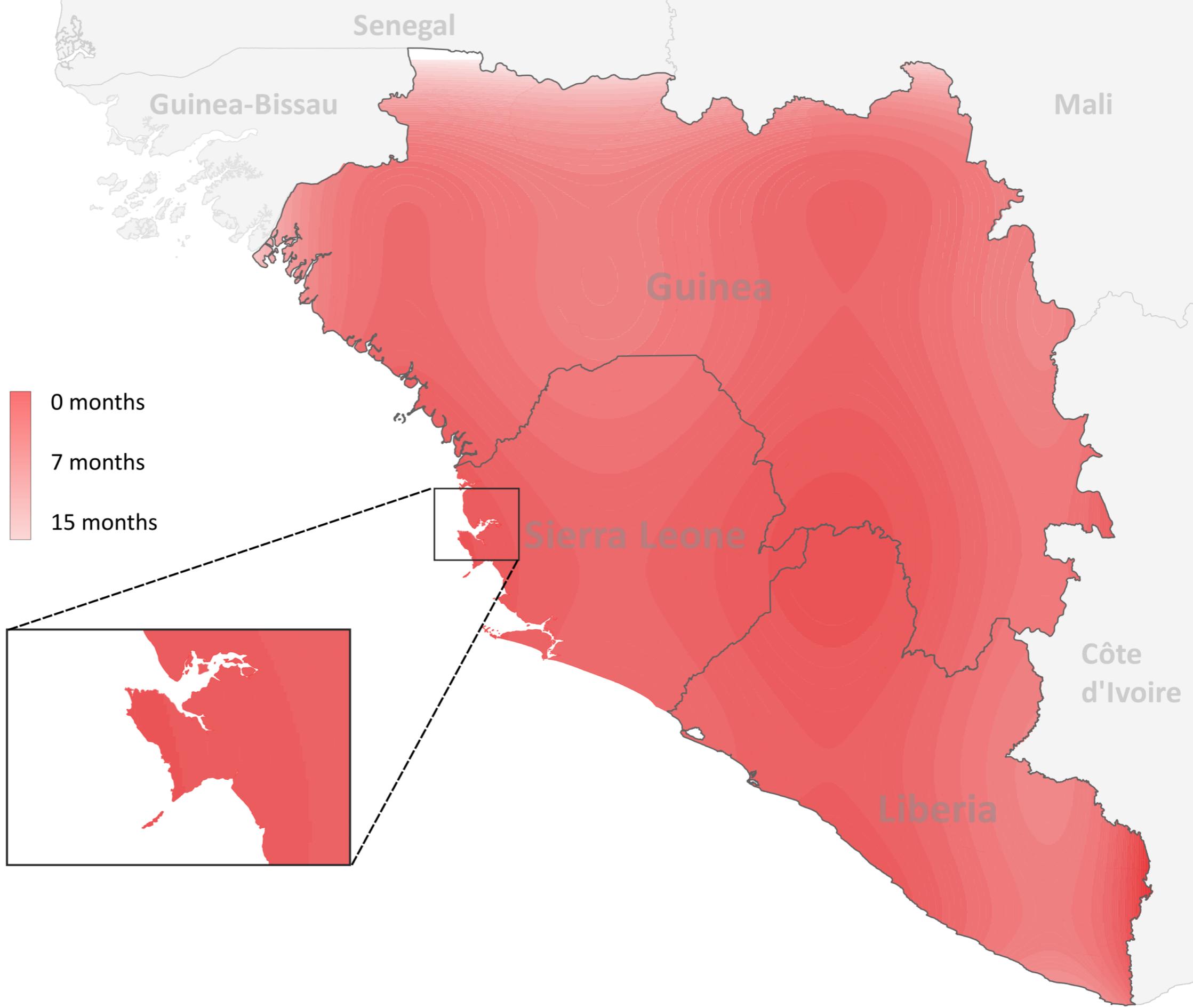
Methods

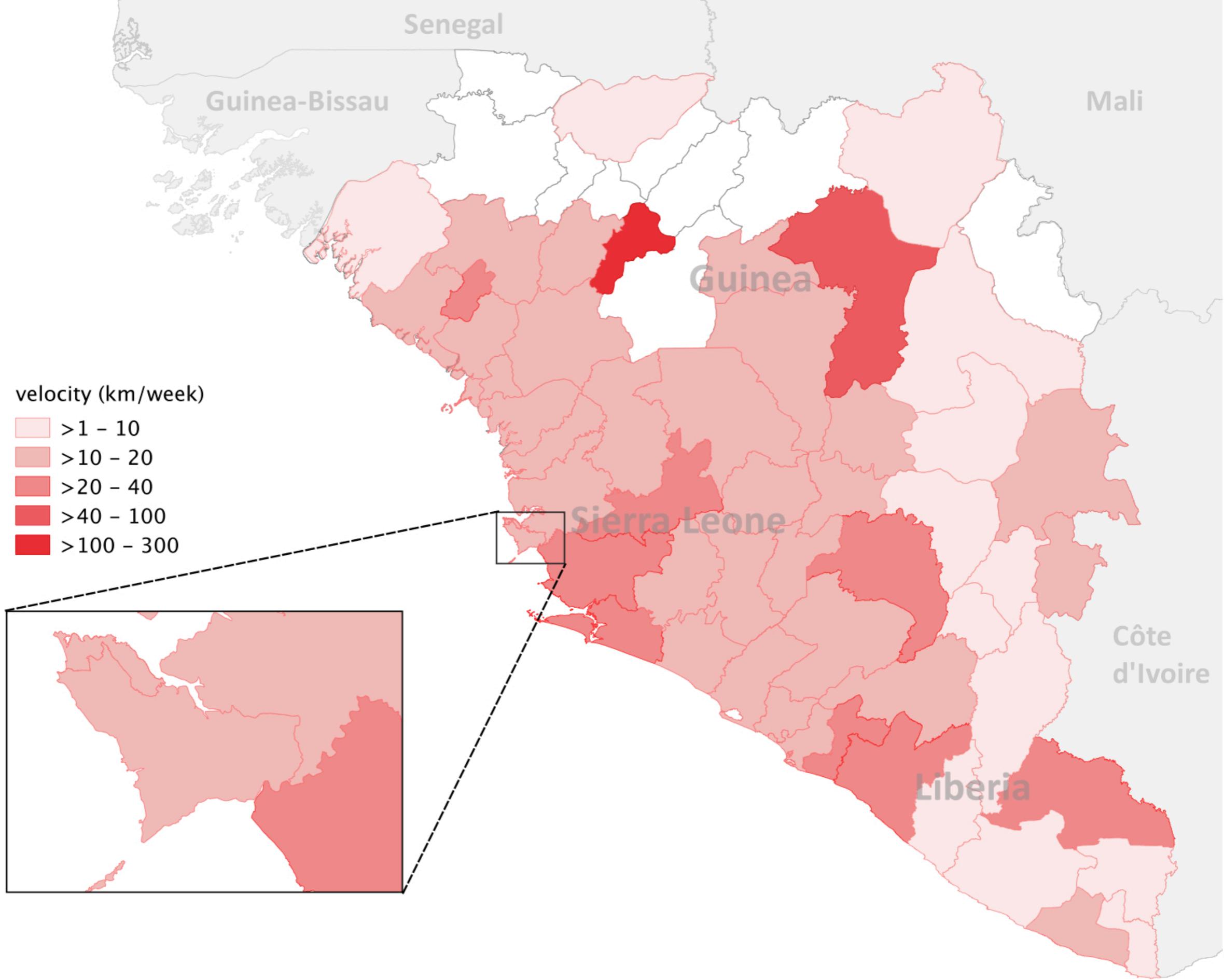
- Front-wave velocity, surface trend analysis
- Spatial interpolation method based on least squares regression to study diffusion in space and time
- Surface pattern constructed by mapping the specific timing of events at each (x,y) coordinate
- Shape and flexibility are determined by the polynomial order chosen (increasing order increases local curvature of fitted surface)

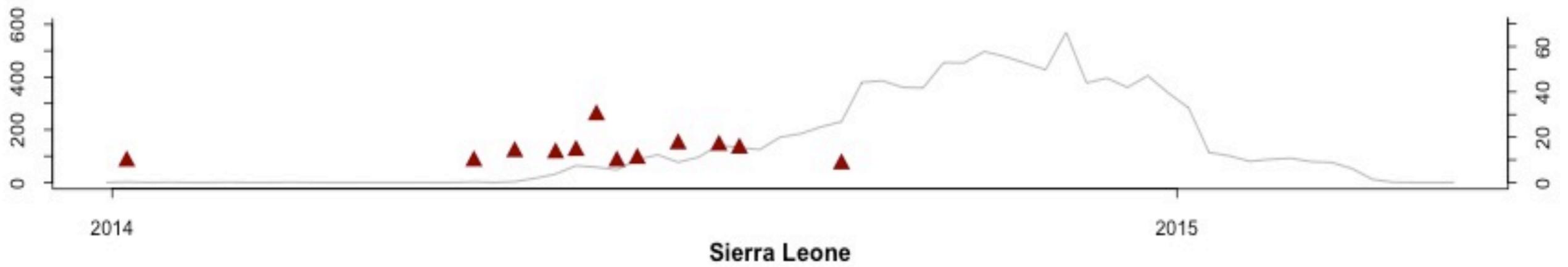
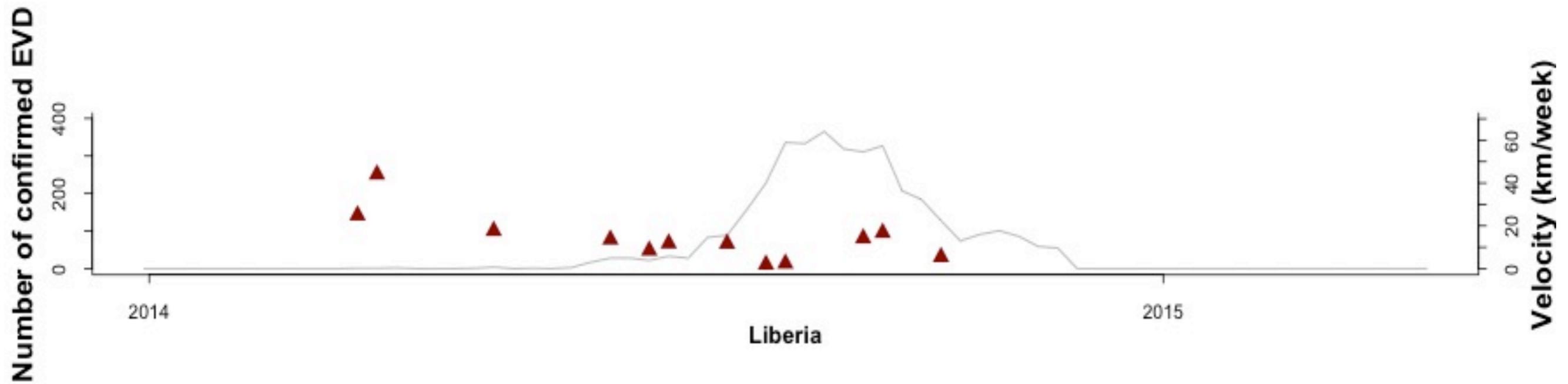
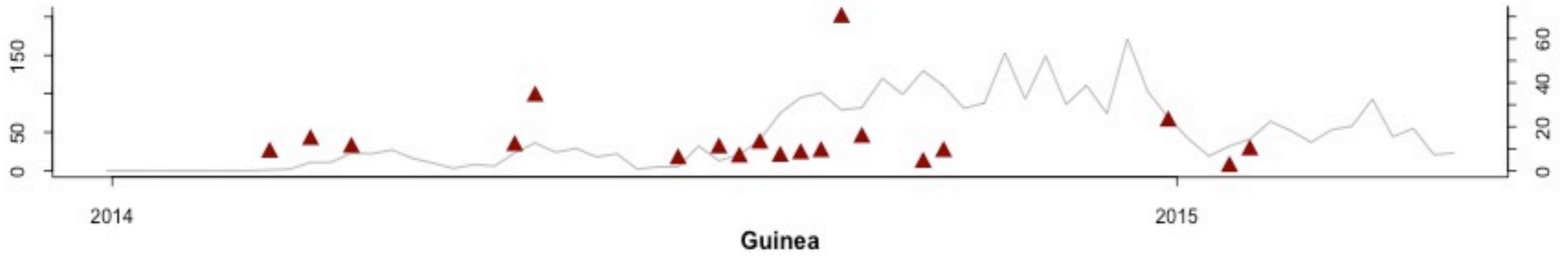
Model

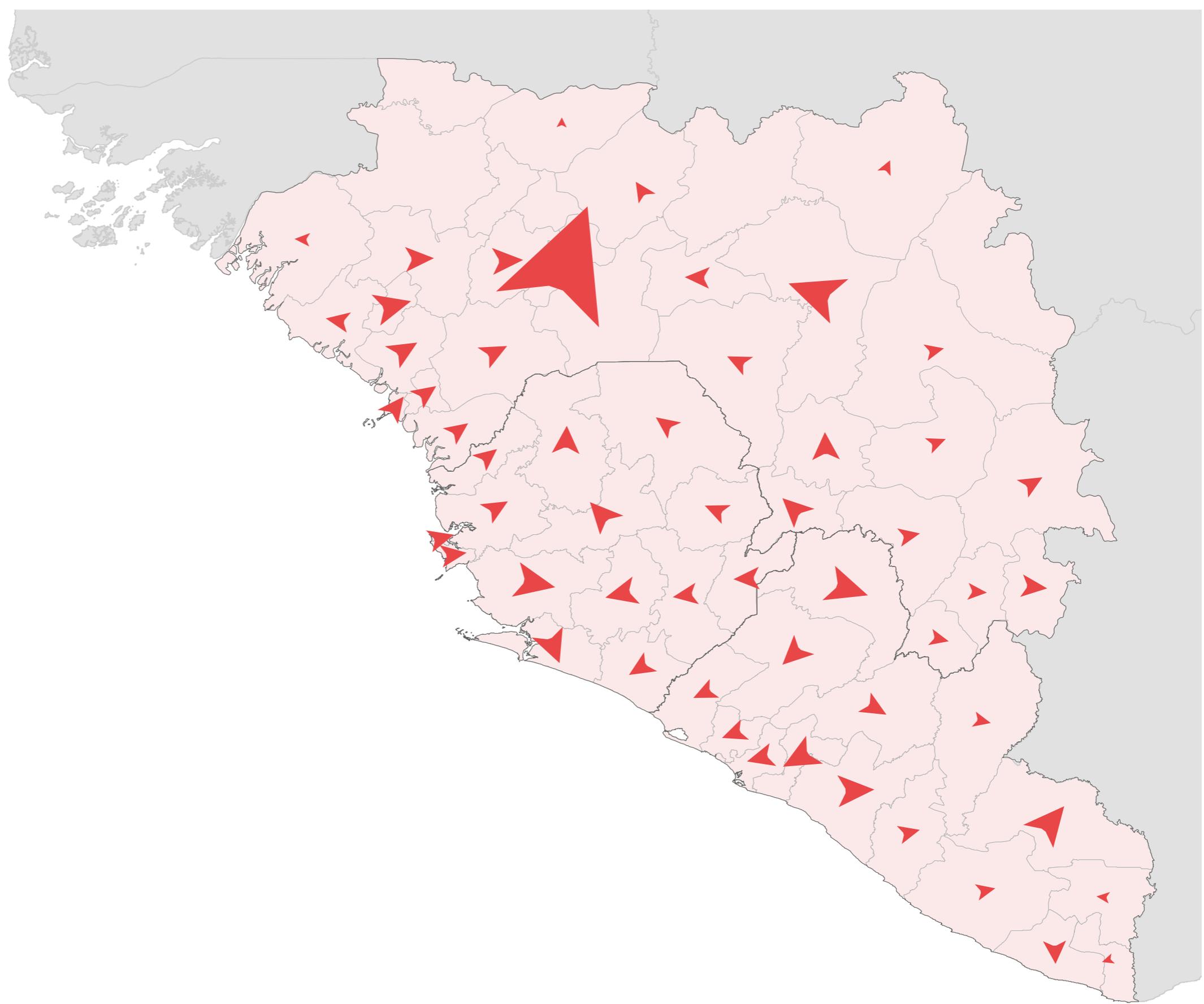
- $f(t|x,y) = \beta_0 + \beta_1X + \beta_2Y + \beta_3X^2 + \beta_4Y^2 + \beta_5X^3 + \beta_6Y^3 + \beta_7X^4 + \beta_8Y^4 + \beta_9X^5 + \beta_{10}Y^5 + \beta_{11}X^6 + \beta_{12}Y^6 + \beta_{13}XY$ (1)
- $\partial f(t|x,y) / \partial x = \beta_1 + 2\beta_3X + 3\beta_5X^2 + 4\beta_7X^3 + 5\beta_9X^4 + 6\beta_{11}X^5 + \beta_{13}Y$ (2)
- $\partial f(t|x,y) / \partial y = \beta_2 + 2\beta_4Y + 3\beta_6Y^2 + 4\beta_8Y^3 + 5\beta_{10}Y^4 + 6\beta_{12}Y^5 + \beta_{13}X$ (3)
- Partial derivatives (2) and (3) provide expressions for a slope vector at a given location X, Y
- Magnitude and direction of rate of change (in weeks per km) was estimated by the inner product of the vector, where magnitude $\|xy\| = \sqrt{(x^2 + y^2)}$ and the direction $\theta = \tan^{-1}(y/x)$
- Residuals assessed for spatial autocorrelation

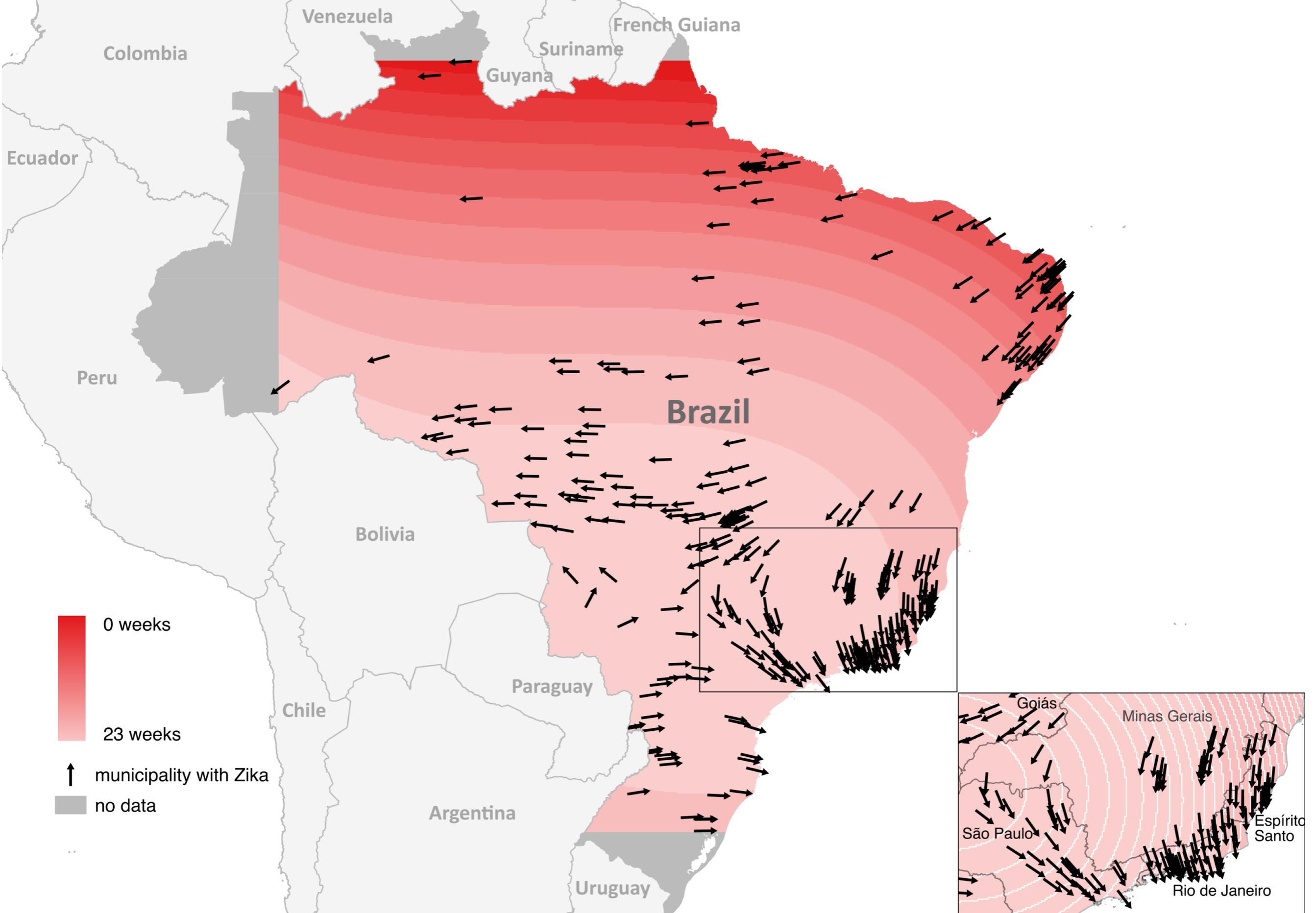












Study limitations

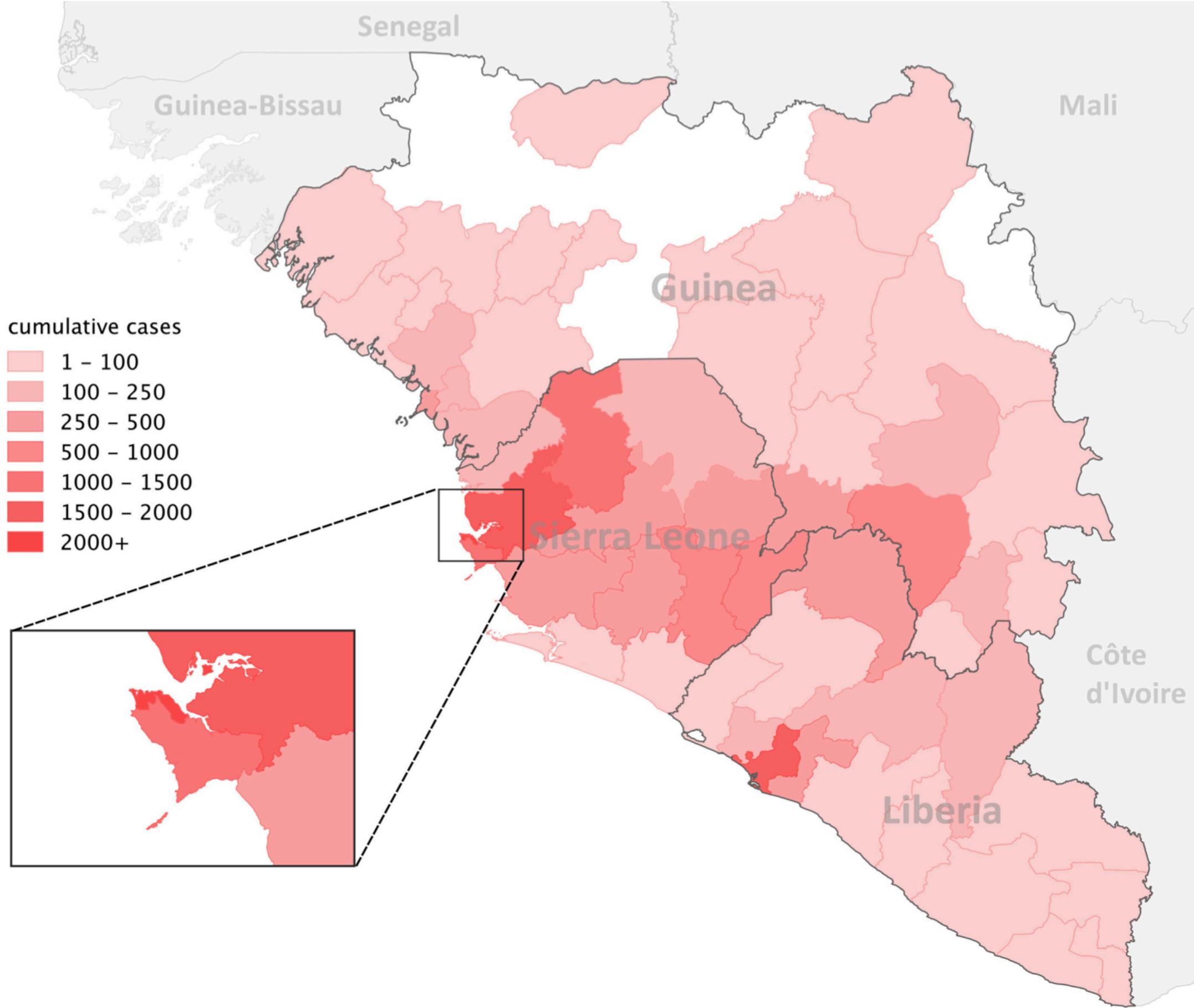
- WHO data subject to error including accuracy and completeness
- Week of first confirmed EVD case may have been based on different dates including date of symptom onset or date of laboratory report
- Could have been earlier introductions of EVD that were not captured by surveillance system
- Data based on large geographic areas, increasing uncertainty associated with velocity estimates including edge effects

Conclusions

- Predicting disease movement is useful for identifying critical locations and corridors for containment efforts
- These methods should be applied to other diseases
- Extended to prediction and other predictors

Next analysis...understanding why

- Predictors of Ebola ecological niches, spillover events, or the onset of EVD outbreaks
 - EVD-related deaths in primates, deforestation, human forest activities, and population density
 - Precipitation, humidity, transition seasons, elevation, temperature, vegetation and vegetation density
- Several mathematical modelling studies estimate the size, speed, and patterns
- Few spatial studies based on empirical data
 - Social vulnerability was associated with spatial EVD transmission
 - Population, cases, and distance was associated with spatial risk of EVD infection



Objective & Methods

- Identify environmental and population-level demographic spatial predictors of human EVD risk
- Potential predictors
 - Environmental data, elevation, land area, roadway and waterway density
 - Population density
 - National DHS – household education, wealth, occupation, household structure, possessions and amenities
- Regressed cumulative total EVD cases on selected covariates
 - Selected model with best AIC
- Quantified amount of variation of EVD cases by using a spatial autoregressive modelling approach

Model

- Two random effects (uncorrelated residual variation and spatially correlated residual variation)
- INLA used to estimate marginal posterior distributions

$$Y_i \sim \text{Poisson}(\mu_i)$$

$$\log\left(\frac{Y_i}{\text{pop}_i}\right) = \alpha_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_{12} x_{12} + u_i + \nu_i$$

$$u_i \sim N\left(0, \frac{1}{\tau_u}\right)$$

$$\nu_i | \tau_\nu, \nu_j, i \neq j, \sim N\left(\frac{1}{n_i} \sum_{i \sim j} \nu_j, \frac{1}{n_i \tau_\nu}\right)$$

$$\alpha_0, \beta_1, \dots, \beta_j \sim N(0, 1000)$$

where i and j refer to two distinct regions in the study area, $i \sim j$ refers to two neighboring regions, and n_i refers to the number of neighboring regions for region i .

Results

Table 1: Country-level summaries for selected covariates

Covariate	Guinea	Liberia	Sierra Leone
Total confirmed EVD cases	3,144	3,339	9,394
Average rainfall accumulation (cm)	3.2	3.6	4.8
Average elevation (m)	435.3	172.1	166.4
Roadway density (km/100 km ²)	10.0	11.2	16.8
Waterway density (km/100 km ²)	11.2	9.0	15.7
Cropland (%)	9.5%	5.1%	13.3%
Female headed households (%)	17.3%	35.2%	28.0%
Secondary education (%)	1.5%	10.5%	4.8%
Households (%) without toilet facilities	19.5%	45.2%	21.4%
Households (%) with electricity	26.2%	9.8%	13.5%
Households (%) with radio	61.5%	58.9%	58.8%

Results

Covariates	Rate ratio (95% credible interval)
Weekly rainfall accumulation (cm)	
<3.22	1.00
3.22-4.18	2.18 (0.66, 7.20)
>4.18	5.34 (1.20, 23.90)
Roadway density	
<0.09	1.00
0.09-0.11	0.61 (0.19, 1.96)
>0.11	1.22 (0.35, 4.26)
Population density	
<33.6	1.00
33.6-68.0	0.64 (0.18, 2.23)
>68.0	0.98 (0.22, 4.35)
Urban land cover (%)	
<0.02	1.00
0.02-0.09	4.87 (1.56, 15.40)
>0.09	5.74 (1.68, 19.67)
Household not possessing radios (%)	
<38.1	1.00
38.1-47.6	2.79 (0.90, 8.78)
>47.6	4.23 (1.16, 15.93)

Discussion

- Lack of radio ownership a strong predictor of EVD risk at district level
 - Serial dramas and popular music used to disseminate risk communication
 - Requires further study
- Rainfall and EVD transmission risk
 - Roads become impassable with higher rainfall
 - Time series approach may be helpful
- Roadway and waterway densities as proxies for population mobility
 - High roadway density could decrease risk by improving accessibility
 - Artifact of measurement error?
- Slight protective association with population density
- Increased risk in more urban areas could reflect population mobility

Limitations & Conclusions

- Missing important predictors
- Measurement error
 - Remote sensing data
 - DHS estimates at regional level for Guinea
 - Mismatch of time period for different data sources
- Our work has shed new light on population-level factors for EVD transmission risk
 - Future work should examine causality
 - Role of radio requires further study and modifiable risk factor for future outbreaks
- Need higher spatial resolution data and other important predictors

Summary

- Need for quick analysis to inform public health action in EID situations
 - Balance of data and methodological quality
- Need for enhanced TSA applied to different diseases (or other methods?)
- Understand when spatial confounding could be an issue
- Causality...the goal but difficult to achieve with surveillance data
- Need more (bio)statisticians on editorial boards of public health journals

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