

Swiss TPH



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Modeling of Large Geostatistical Data to Estimate Seasonal and Spatial Variation of Sporozoite Rate in Rufiji DSS

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- Introduction
- Data + Methods
- Application + Results
- Discussion + Conclusion

- ❖ Seasonal behaviors of most factors (e.g. climate and environment)
 - ❖ Attribute to spatial and seasonal fluctuation on health outcome
 - ❖ Interfere with the ecological environment of vectors or their behaviors
- ❖ Malaria transmission has shown seasonal variations in many regions in the Sub Saharan Africa.
 - ❖ These change spatial-temporal distribution of the incidence of malaria
- ❖ Understanding the seasonal and geographical variation of malaria transmission is important:
 - ❖ targeted timely control strategy
 - ❖ disease dynamics
 - ❖ prediction of epidemics
 - ❖ better design and evaluation of control and intervention
 - ❖ relate patterns of transmission with other health outcome

- ❖ A number of entomological studies have looked at small area variations of the malaria transmission.

- ❖ Very few were based on entomological data
 - ❖ collected over a long period of time
 - ❖ with high temporal resolution

- ❖ No studies looking at spatial variation of seasonality using ento. data with high spatio-temporal scales.
 - ❖ **Difficult to draw general conclusion**

- ❖ The INDEPTH-Malaria Transmission Intensity and Mortality Burdens across Africa (MTIMBA) project (2001-2004)
 - ❖ initiated to study the relation between malaria transmission and mortality.
 - ❖ data were collected longitudinally
 - ❖ at a large number of locations every 2 weeks
 - ❖ in a systematic and standardized manner.

- ❖ Data are correlated in space and time
 - ❖ Leading to geostatistical models having large number of parameters.
 - ❖ Model fit is most appropriately done using Bayesian MCMC simulations.

- ❖ Several methods have been proposed to estimate the spatial processes of large geostatistical data.
- ❖ This study develops the space-time model by further extend existing methodology which approximate the spatial process from a subset of locations and,
- ❖ Apply to analyze MTIMBA data from the Rufiji DSS.
 - ❖ Study the seasonality of the SR of the two species, *An. funestus* and *An. gambiae*.
 - ❖ Prediction of the SR for the whole DSS area using results obtained from the approximation methods.

Outcome

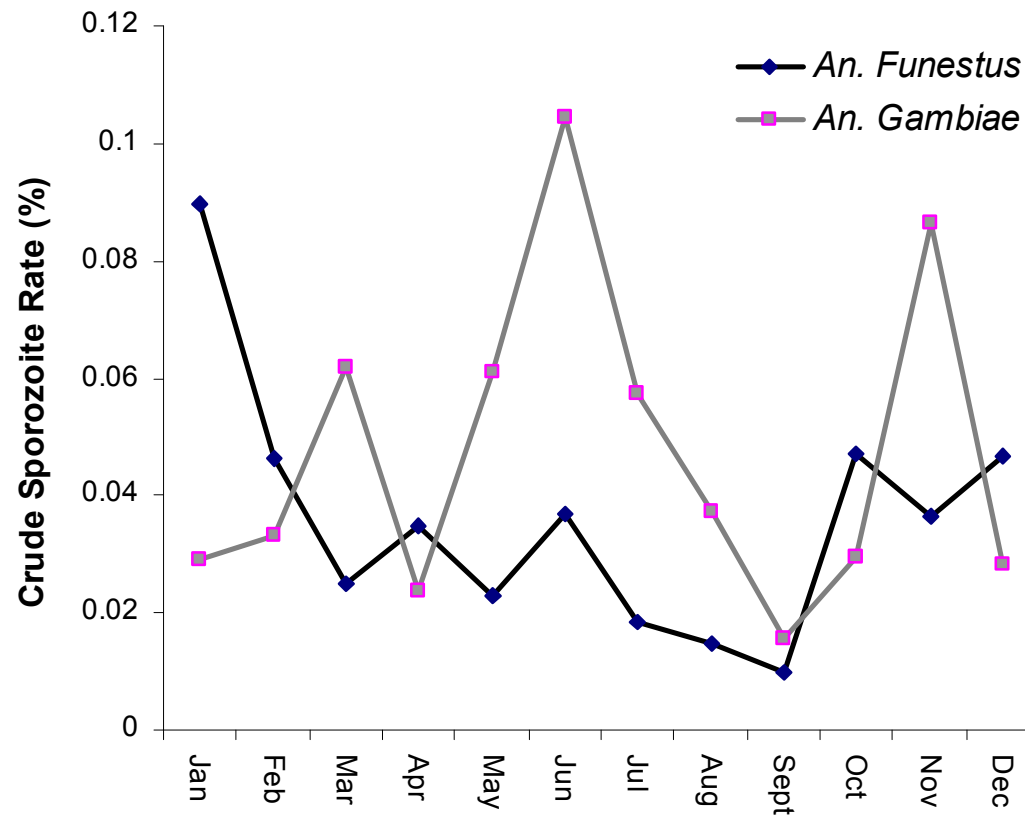
- ❖ 16,325 (393 positive; SR=2.41%) *An. funestus* mosquitoes
- ❖ 10,610 (450 positive; SR=4.24%) *An. gambiae* mosquitoes
 - ❖ Data were summarized by locations and calendar months.
 - ❖ Repeated surveys (locations) within the same month were collapsed to a single observation.

- ❖ 430 data points for *An. funestus* (415 unique locations).
- ❖ 670 data points for *An. gambiae* (639 unique locations).

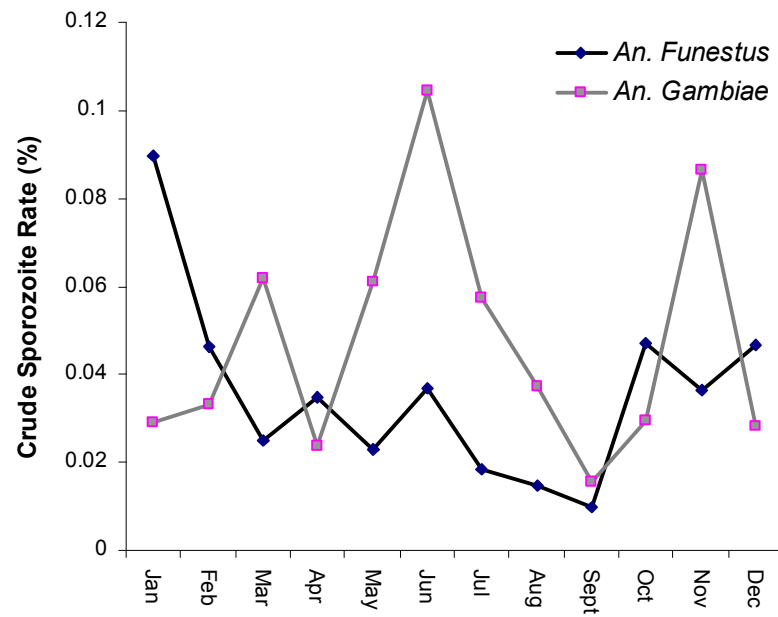
Predictors

- ❖ normalized difference vegetation index (NDVI)
- ❖ Temperature (day)
- ❖ rainfall
- ❖ distance to water bodies.

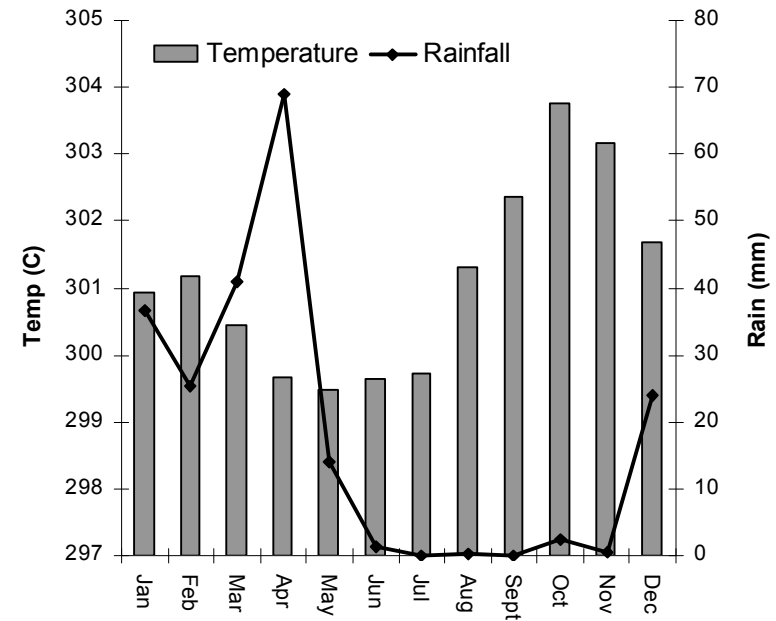
Crude monthly sporozoite rate for *An. funestus* and *An. gambiae*



Crude monthly sporozoite rate



Rain and Temperature



Selection of knots (sub-locations)

- ❖ Use balance sampling procedure
- ❖ Preserve the variation of the outcome
 - ❖ 100 locations out of 415 for *An. Funestus*
 - ❖ 250 locations out of 639 for *An. Gambiae*

Capturing seasonality

- ❖ Use harmonic cycles of cosine function with a mixture of two periods, 6 and 12 months
- ❖ Programs implemented in FORTRAN

Maps of SR for both species

Spatial and Seasonal differences

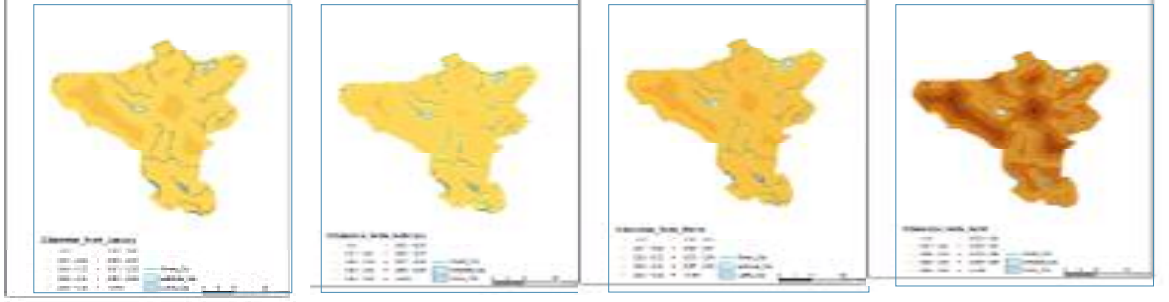
Overall trend of the SR (predicted)

An. funestus

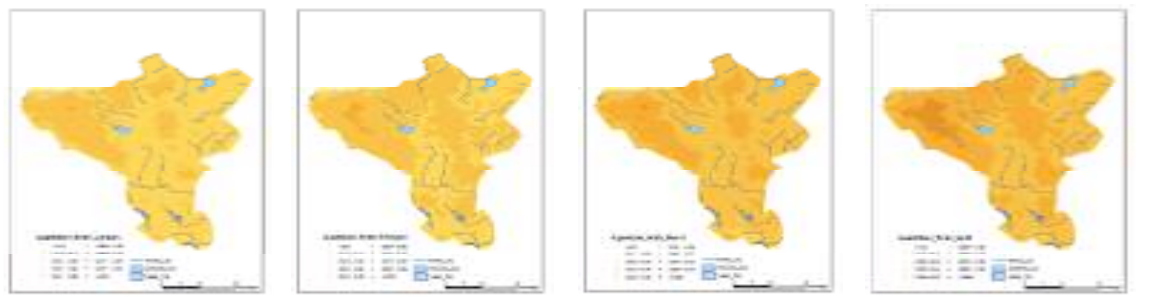
An. gambiae

Monthly SR for *An. funestus* and *An. gambiae*

An. funestus

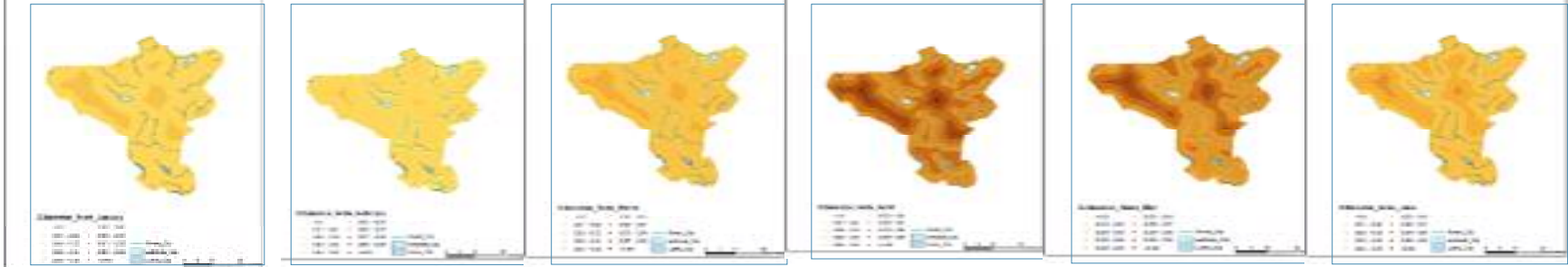


An. gambiae

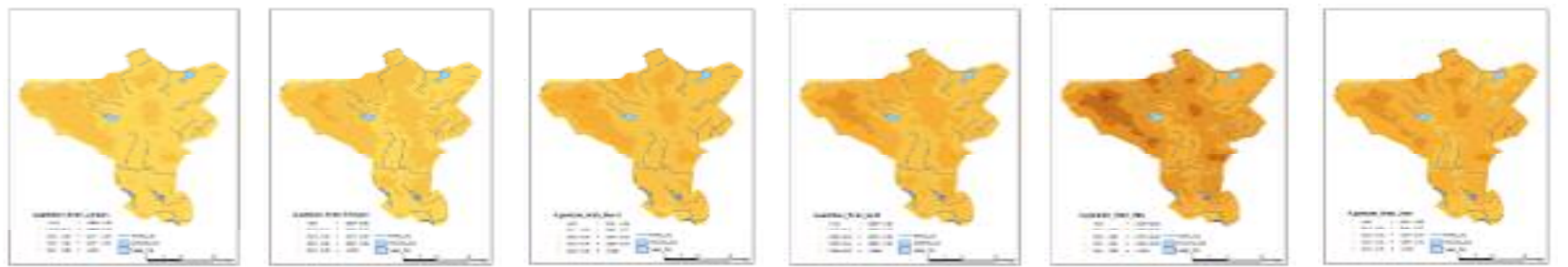


Monthly SR for *An. funestus* and *An. gambiae*

An. funestus

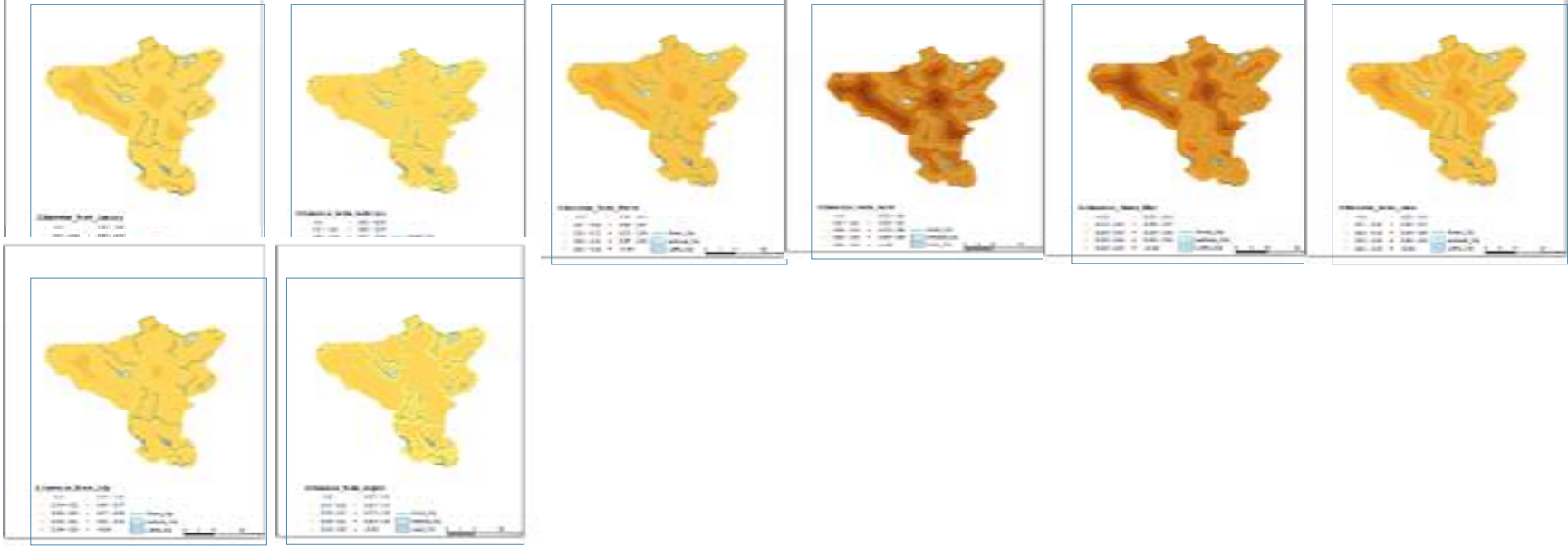


An. gambiae

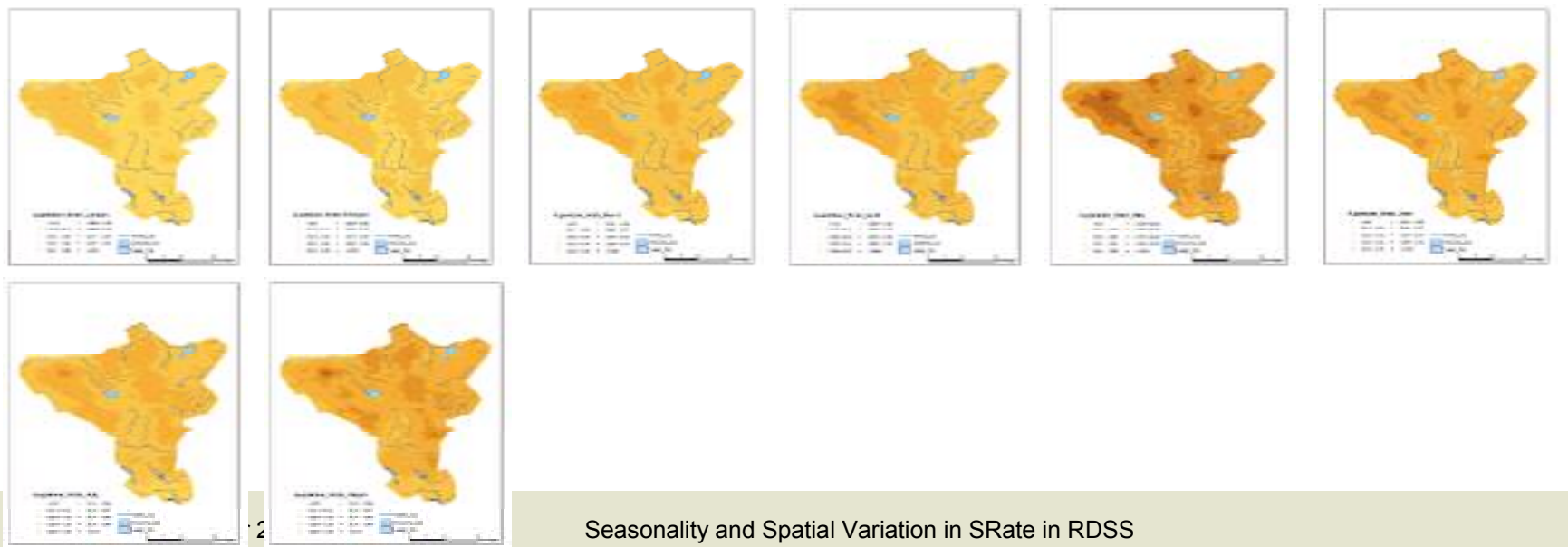


Monthly SR for *An. funestus* and *An. gambiae*

An. funestus

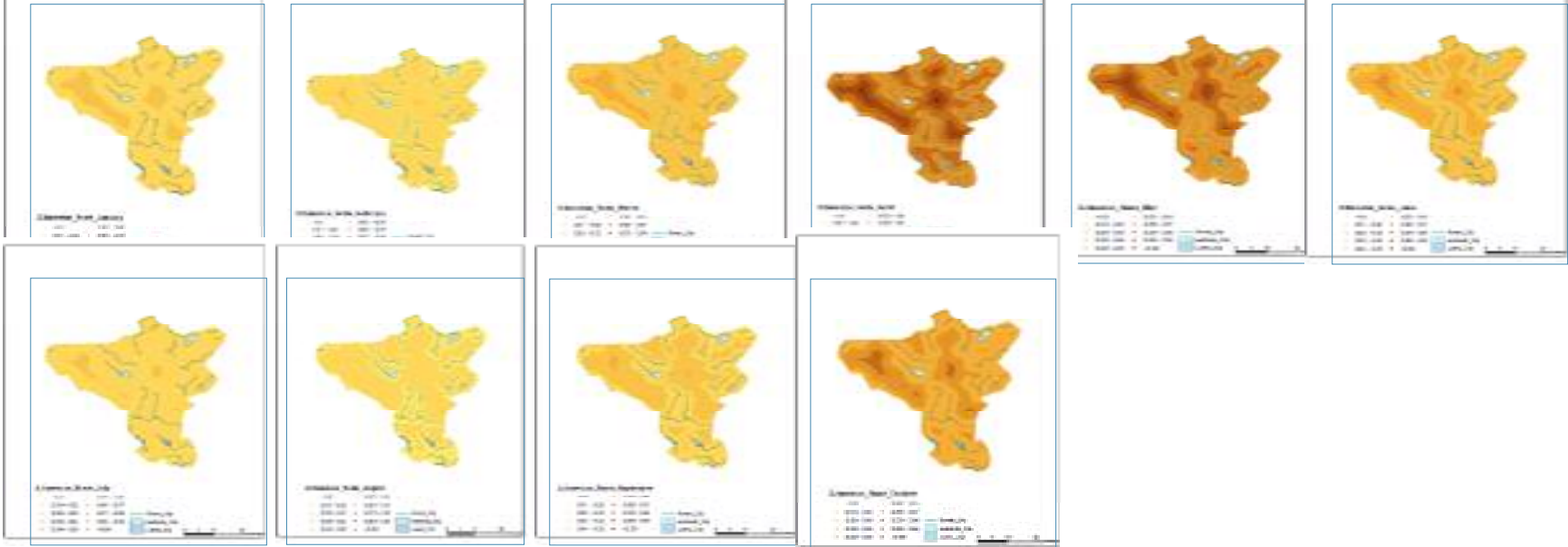


An. gambiae

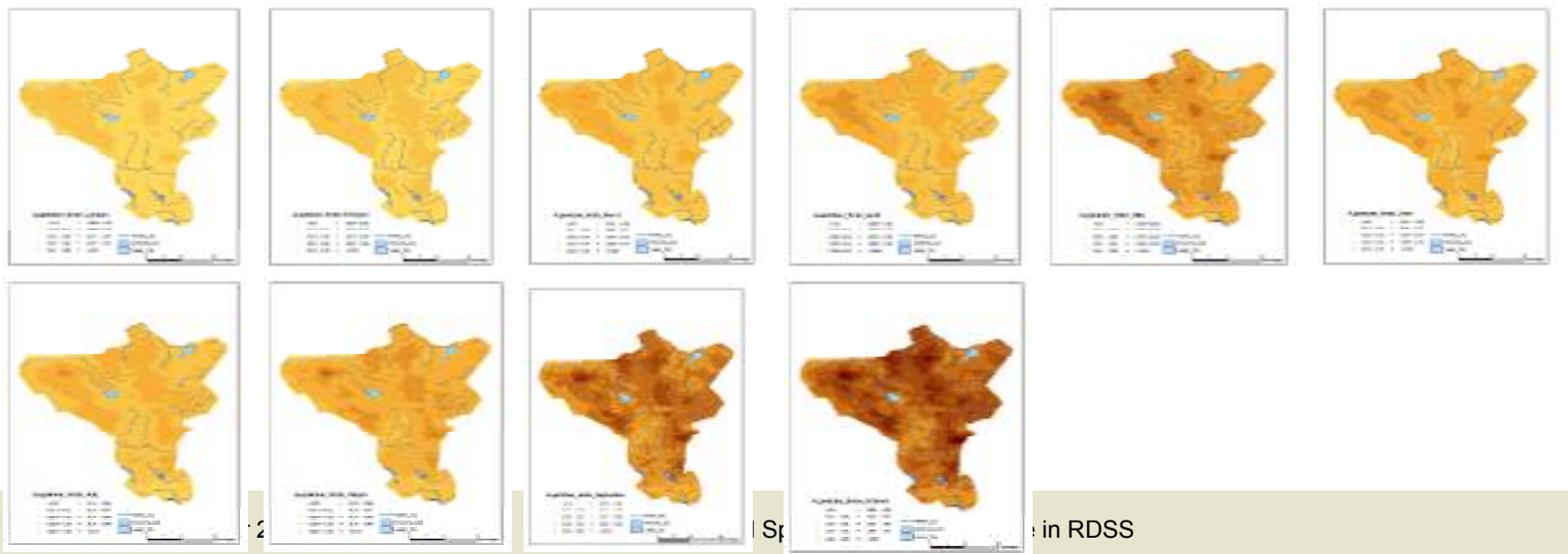


Monthly SR for *An. funestus* and *An. gambiae*

An. funestus

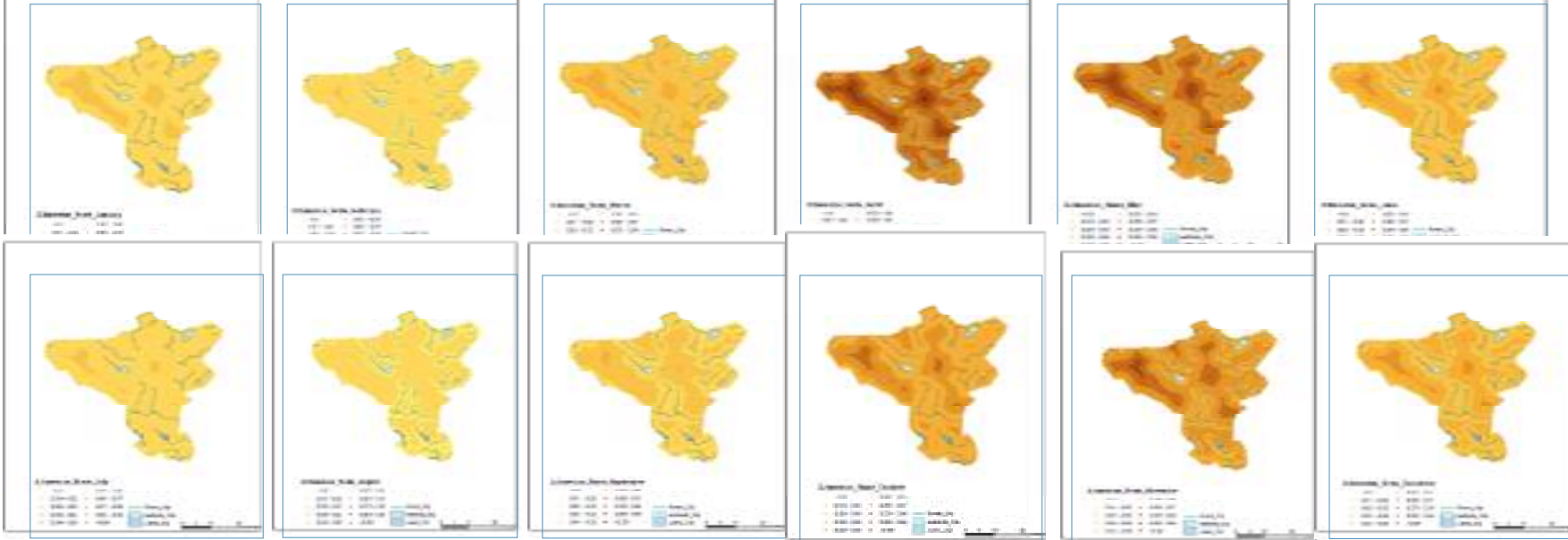


An. gambiae

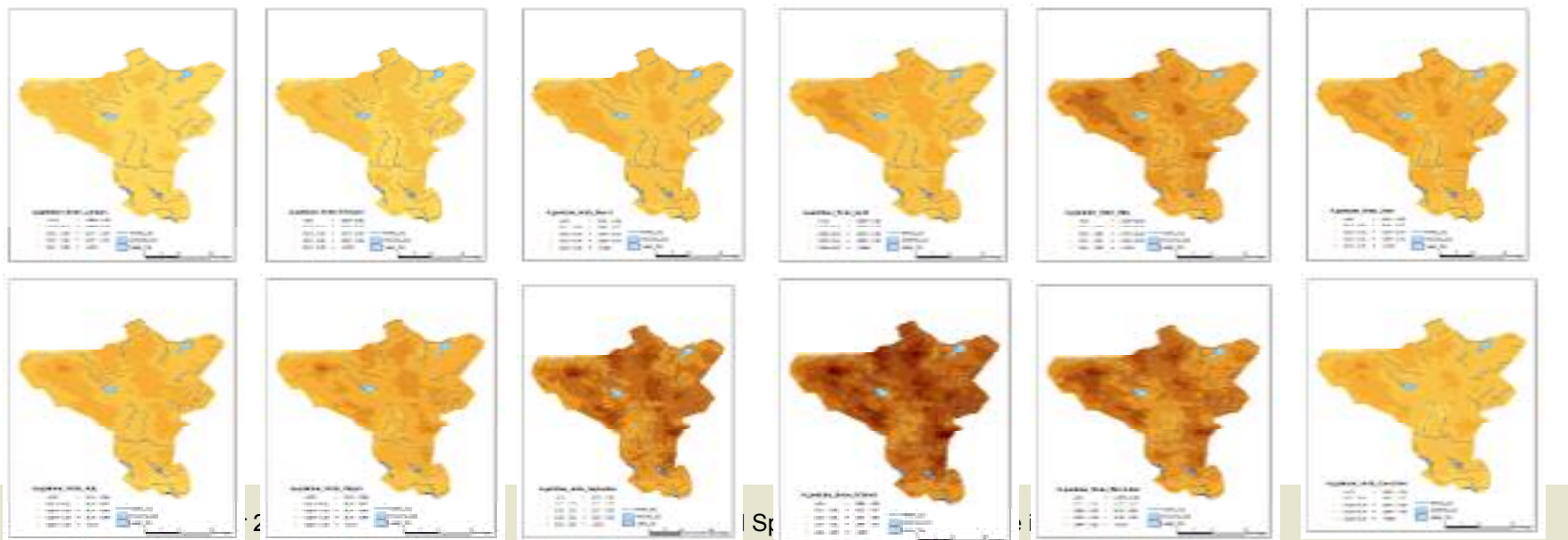


Monthly SR for *An. funestus* and *An. gambiae*

An. funestus



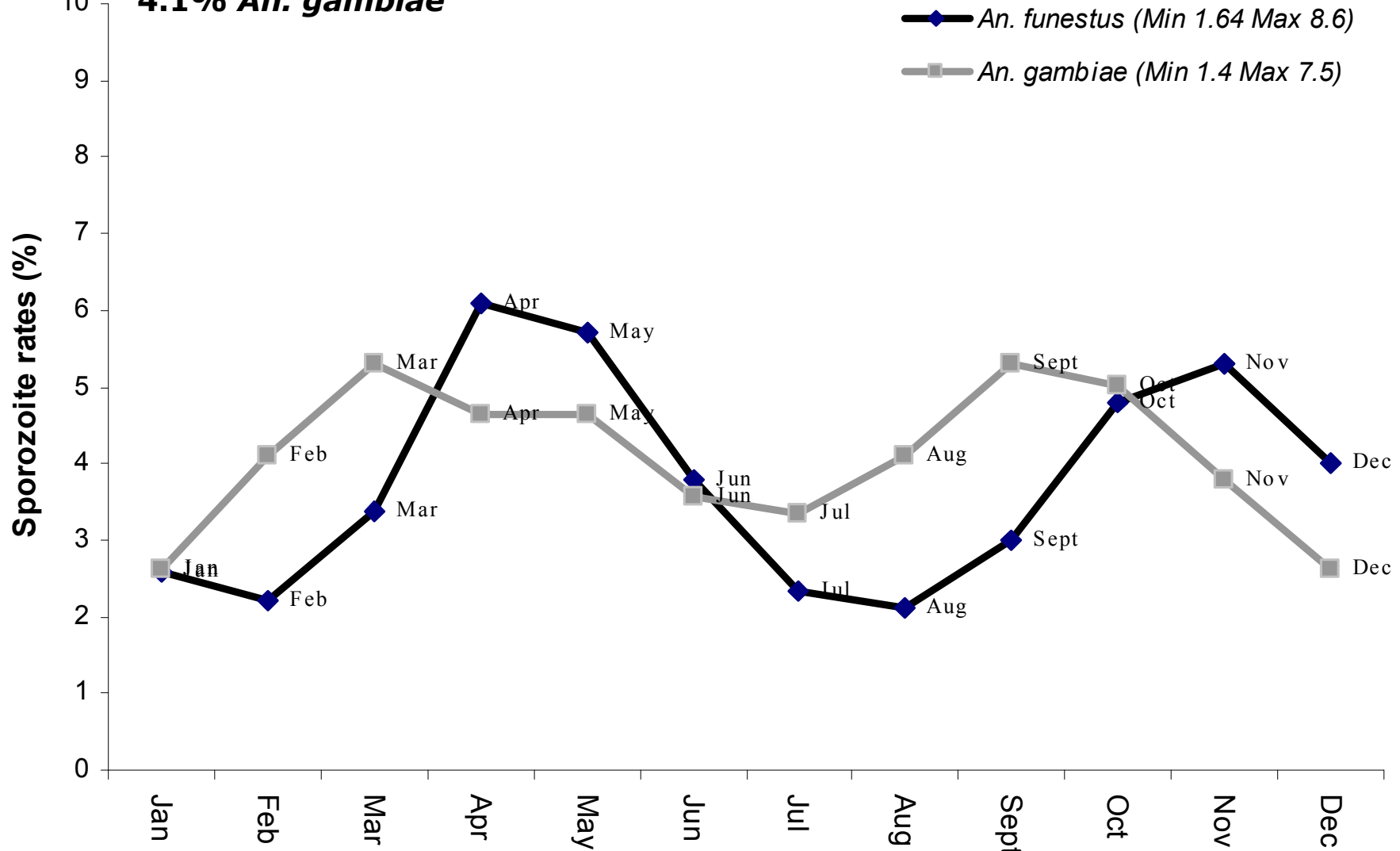
An. gambiae



- ❖ Significant negative effect of temperature and rainfall.
- ❖ The spatial correlation was still significant (more than 5%) till a distance of about 9km.
 - ❖ This means up to 9kms the SR within locations are still correlated.

Mean predicted sporozoite rates

Annual predicted SR:
3.7% *An. funestus*
4.1% *An. gambiae*



- ❖ We demonstrate a need to generate spatial and temporal data on transmission intensity to guide targeted control of malaria operations
- ❖ As a part of the ongoing control and monitoring strategies:
 - ❖ Close follow up changes on climate and environment + occurrence and reoccurrence of breeding sites
 - ❖ Prioritize strategies to locations and periods where transmission peaks
- ❖ This methodology is not only applicable in entomological data but also in analysis of other spatial-temporal disease data with seasonal characteristics.

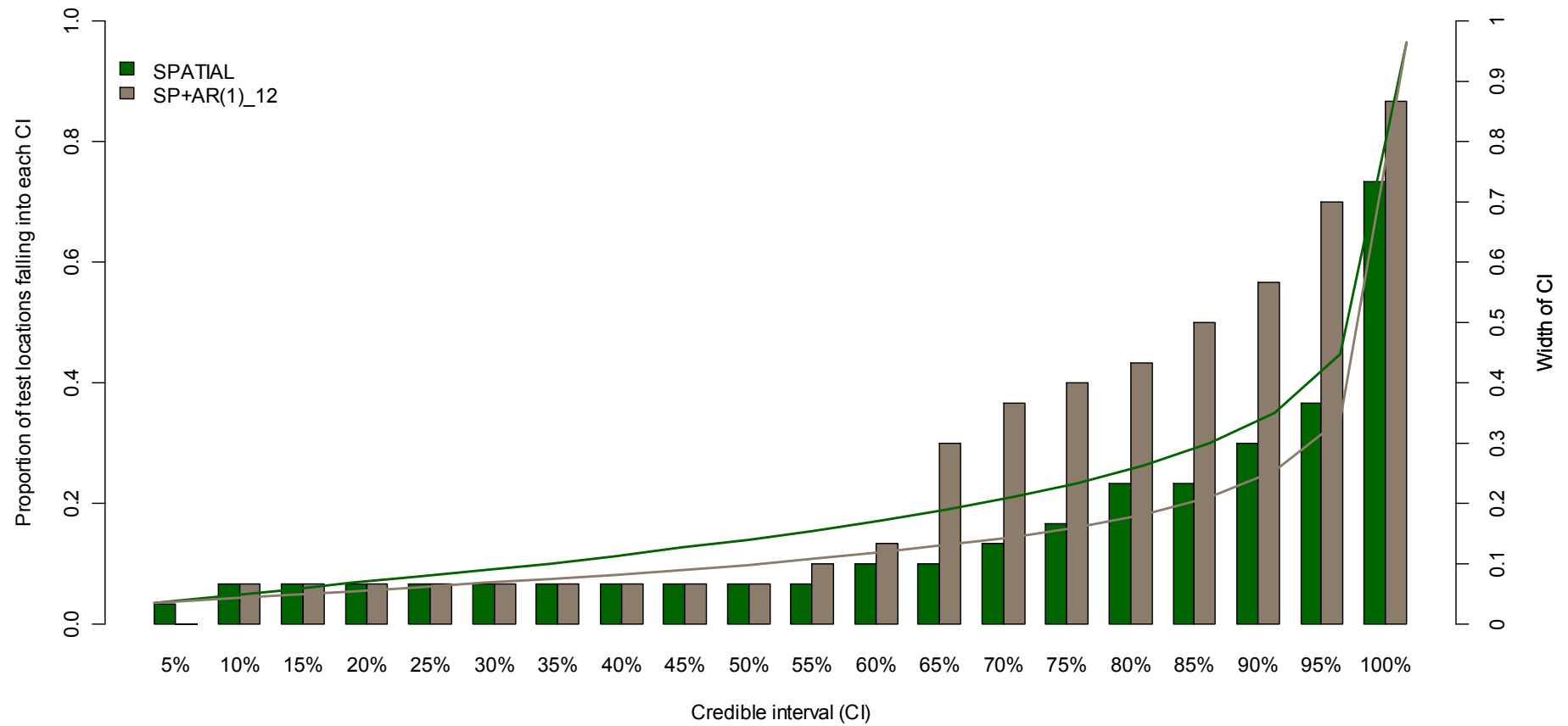
We would like to thank the following:

- ❖ INDEPTH Network team for facilitating the MTIMBA project
- ❖ The Ifakara Health Institute for the management of the project and data.
- ❖ All field workers, data managers and supervisors in the Rufiji DSS.
- ❖ Swiss National Science Foundation financial support

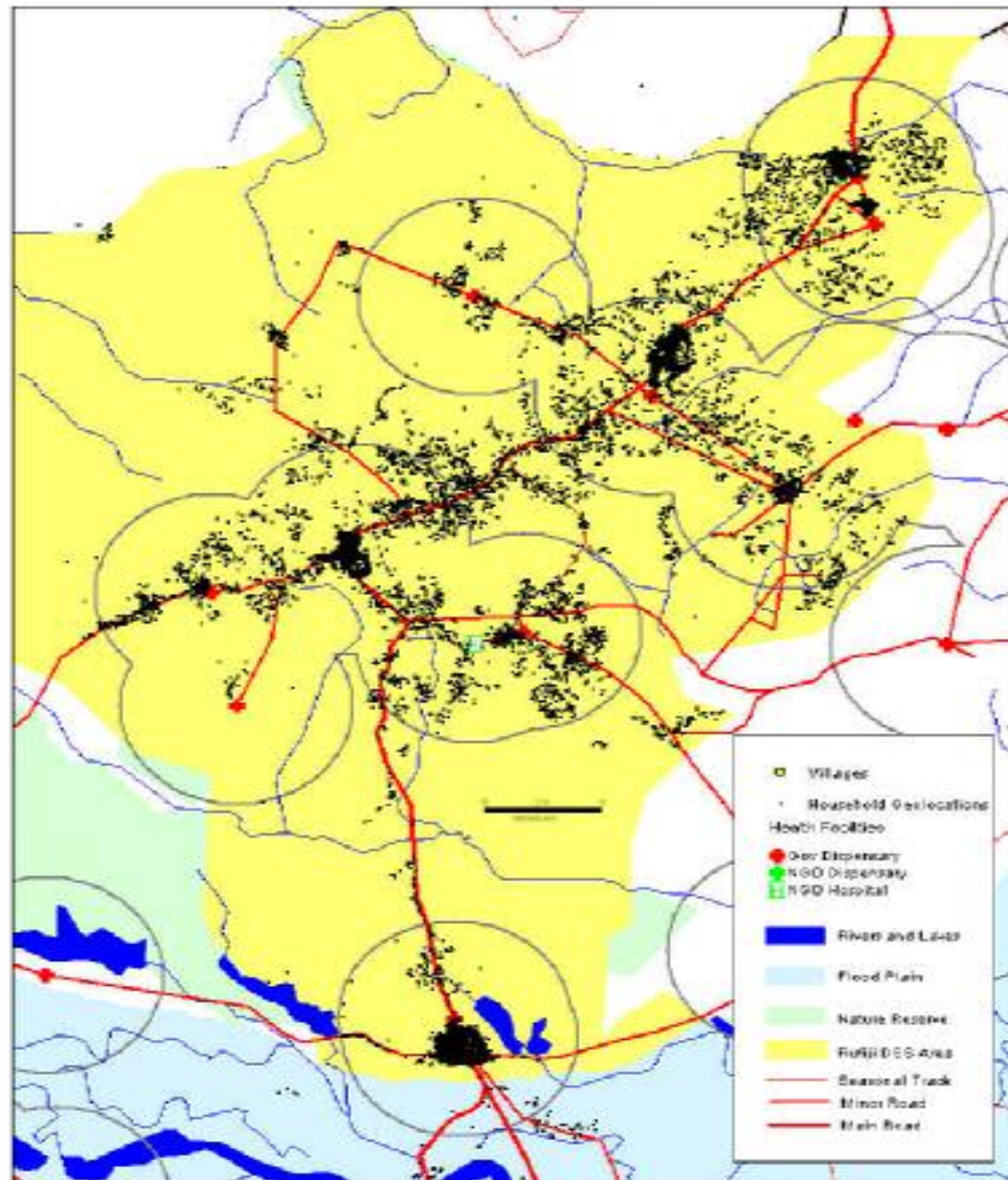
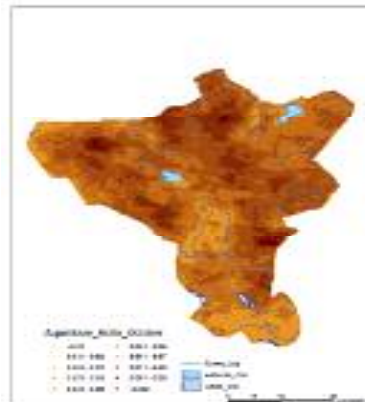
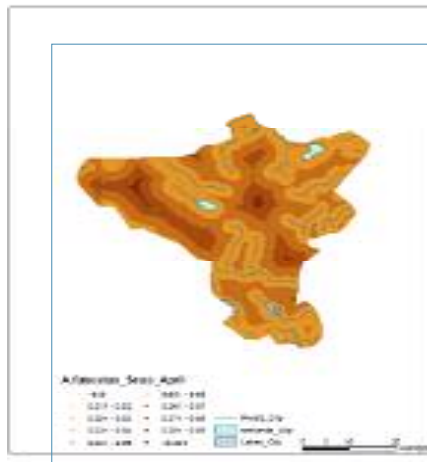
ASANTENI

Model Validation

Bayesian credible intervals



Rufiji DSS Area, Health Facility 5 km Catchments, Tanzania 14,516 Households Registered

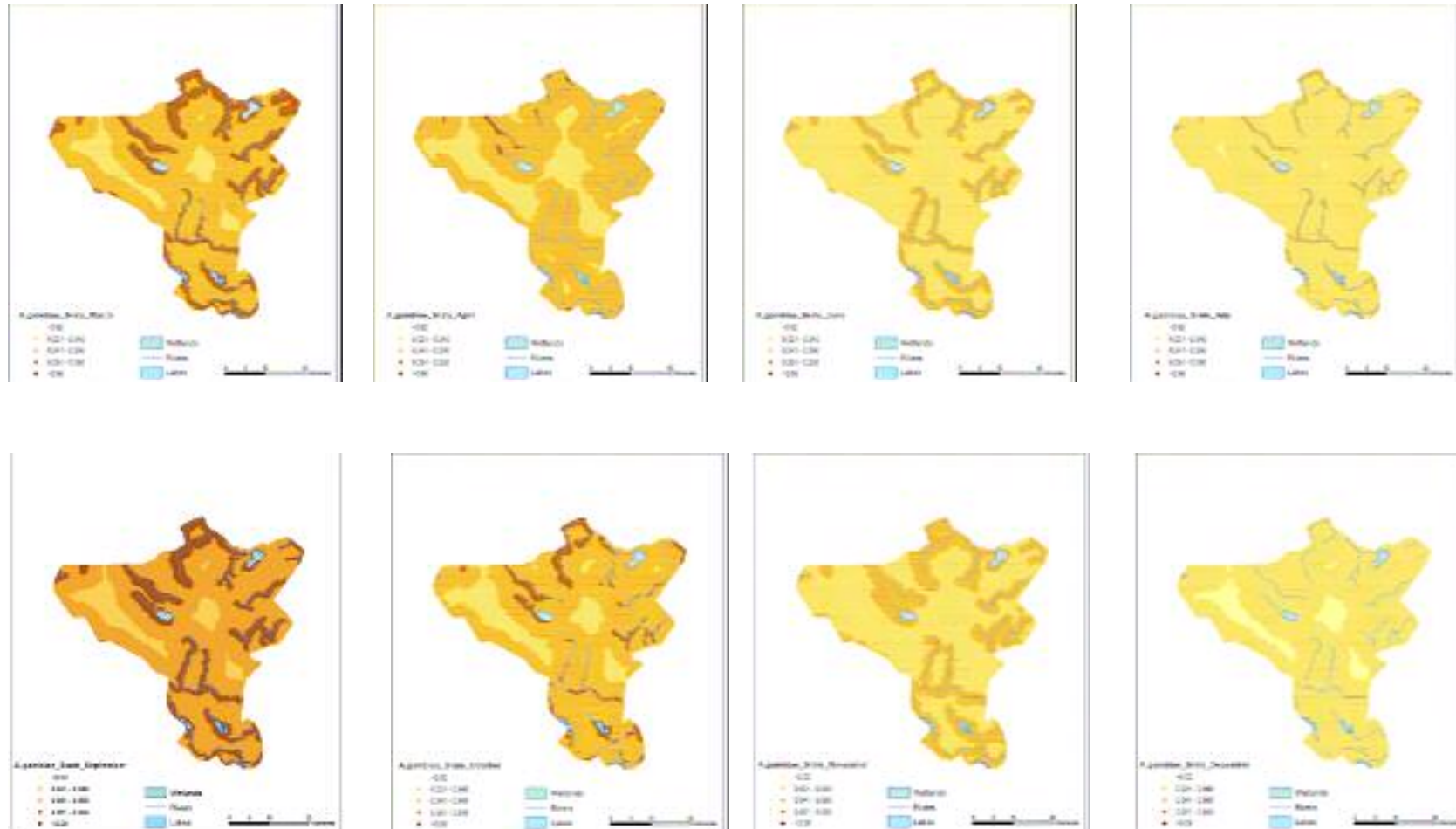


Cartography by TEHIP / MOH, May, 2002

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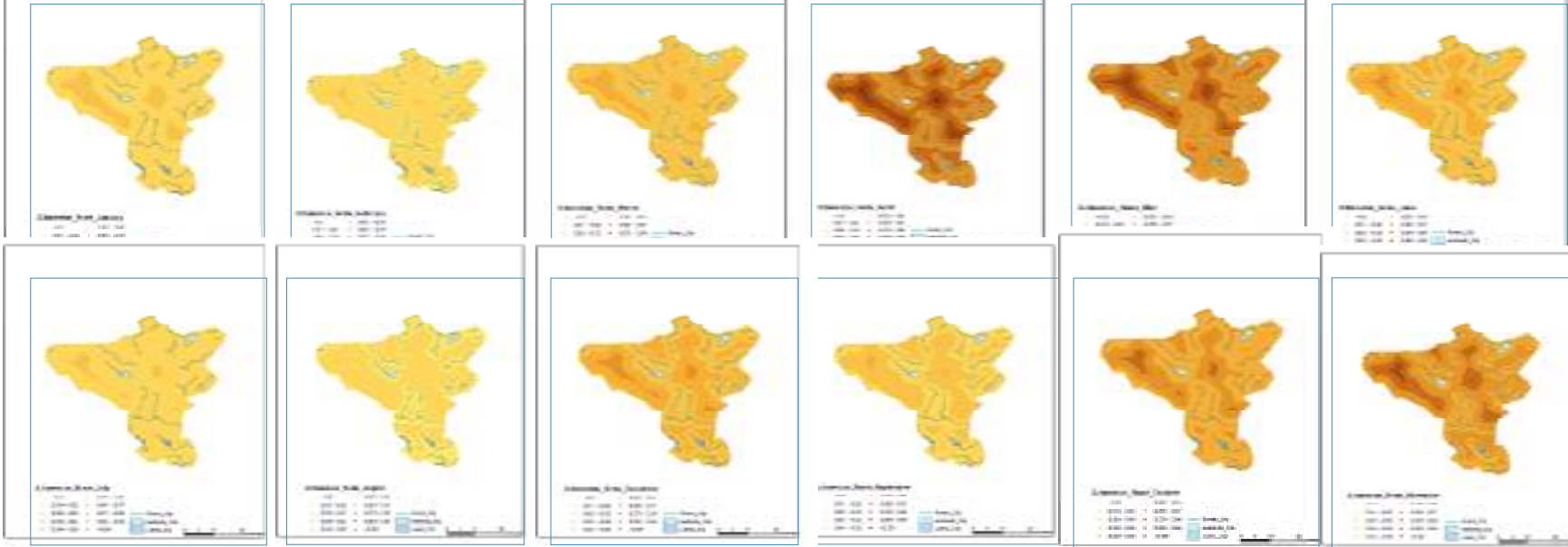
Application + Results



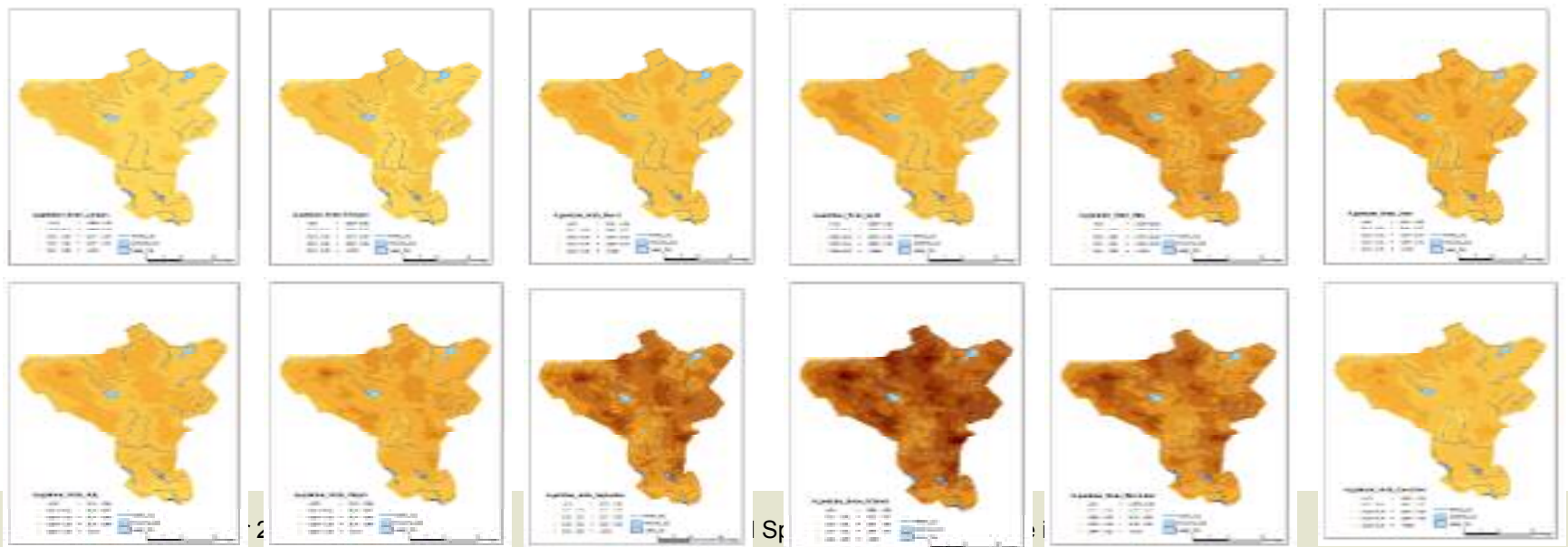
Peaks of transmission are observed during March and September. The heavy and short rains are in March-April and September-October, respectively.

Monthly SR for *An. funestus* and *An. gambiae*

An. funestus



An. gambiae



- ❖ We have used and further develop methodology for modeling large geostatistical data by approximating the spatial process from a subset of locations.
- ❖ The model was able to assess spatial distribution and seasonal pattern of malaria transmission for both species
- ❖ We demonstrates a need to generate spatial and temporal data on transmission intensity on smaller scales to guide targeted control of malaria operations in semi-arid regions

How can this be done:

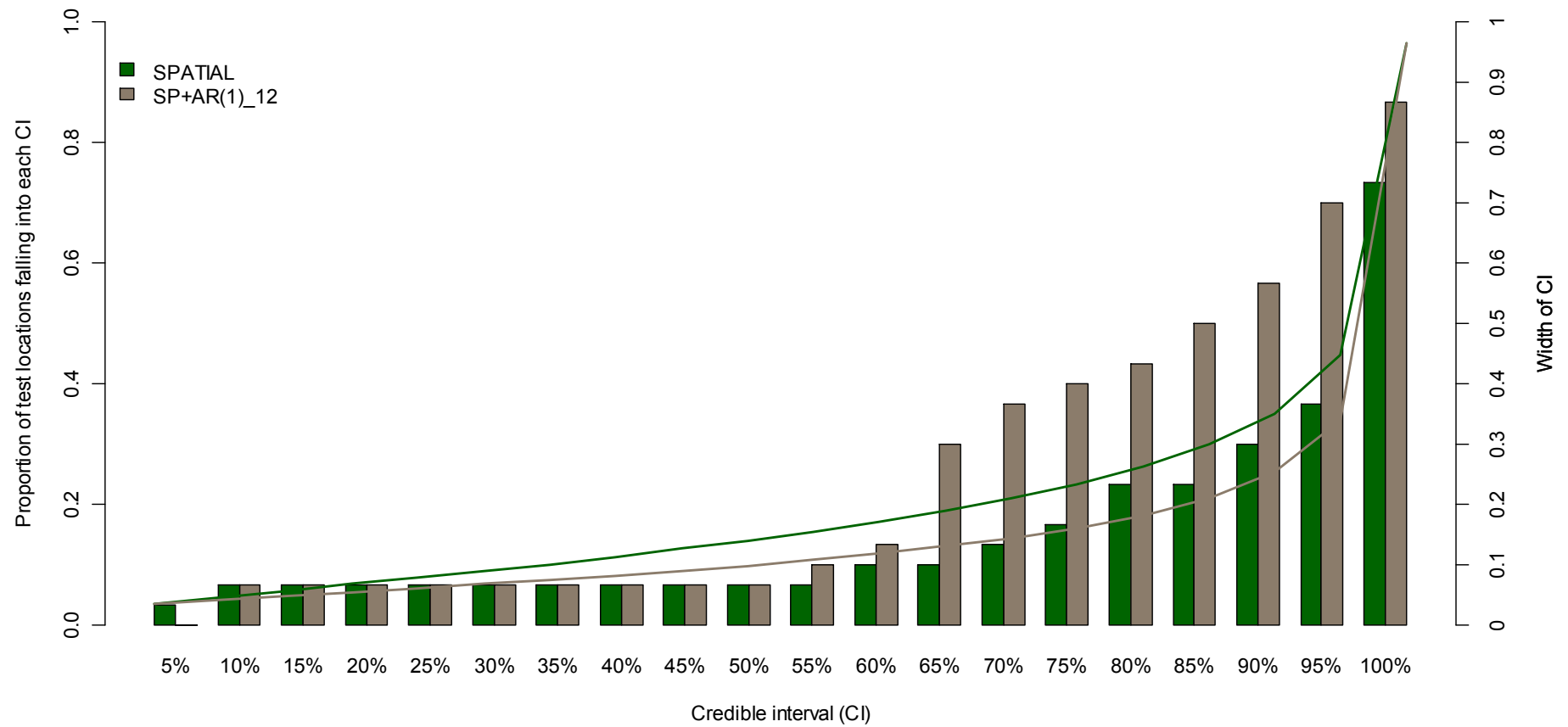
- ❖ As a part of the ongoing control and monitoring strategies:
 - ❖ Close follow up changes on rainfall, temperature and occurrence and reoccurrence of breeding sites
 - ❖ Give priority to locations and periods where transmission peaks to ensure success

- ❖ This methodology is not only applicable in entomological data but also in analysis of other spatial-temporal disease data with seasonal characteristics.

Objectives

Model Validation

Bayesian credible intervals



- ❖ Development of the space-time model and approximate the spatial process using a subset of the locations

- ❖ Application of the method (model)
 - ❖ Study the seasonality of the SR of the two species, *An. funestus* and *An. gambiae*.
 - ❖ Prediction of the SR for the whole DSS area using results obtained from the approximation methods.