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Uma ferramenta para modelagem preditiva de surtos de doenças infecciosas a partir de eventos climáticos extremos

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Global Health Resilience Group | Earth Sciences Department

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Outline

1. Climate-sensitive infectious diseases
2. The IDExtremes project
3. The IDExtremes tool
4. Co-design process and user engagement



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Climate-Sensitive Infectious Diseases (CSIDs)

Diseases that are influenced by, and often exacerbated or modified by, climate factors. Climate-sensitive diseases are typically diseases sensitive to changes in temperature, precipitation, humidity, and other environmental variables associated with climate patterns, such as:

1. **Vector-borne Diseases** (e.g., Dengue, Malaria)
2. **Waterborne Diseases** (e.g., Cholera, Amoebiasis)
3. **Foodborne Diseases** (e.g., Salmonellosis, *E.coli*)
4. **Respiratory Diseases** (e.g., Influenza, COVID-19)

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Over half of known human pathogenic diseases can be aggravated by climate change

Camilo Mora ¹✉, Tristan McKenzie ^{2,3}, Isabella M. Gaw ⁴, Jacqueline M. Dean ¹, Hannah von Hammerstein¹, Tabatha A. Knudson ¹, Renee O. Setter ¹, Charlotte Z. Smith ⁵, Kira M. Webster¹, Jonathan A. Patz⁶ and Erik C. Franklin ^{1,7}

CSIDs and extreme climatic events

Some CSIDs might show a particular sensitivity to extreme climatic events, including heavy rainfall, floods, tropical cyclones, heatwaves and drought.

Disease	Heavy Rainfall	Tropical Cyclones	Drought	Flooding	Heatwaves	Multiple Events
Cholera	Low evidence (N=1) Outbreak ⁵³	High agreement, high evidence (N=7) Outbreak ^{51, 54-59}	High agreement, low evidence (N=2) Outbreak ^{46,52}	Medium agreement, medium evidence (N=4) Outbreak ^{41,60} ⁵⁰ noted the floods did not seed the outbreak; outbreaks began during only one out of every 14 floods ⁴⁶	Low evidence (N=1) Outbreak ³⁰	High agreement, medium evidence (N=3) Including drought followed by heavy rains ⁴⁹ and two systematic reviews investigating water-related disasters ^{48, 61}
Dengue	High agreement, medium evidence (N=3) Increased outbreak risk ^{31,42,43}	Low agreement, high evidence (N=5) Outbreak ^{96,73, 103} No outbreak ^{90,100}	High agreement, medium evidence (N=4) Increased outbreak risk ^{42,43,64,99}	Medium agreement, low evidence (N=2) Unclear impacts of flooding on outbreaks Mixed findings, including decreases and increases ⁹⁵ Decreased risk (possibly due to vector control activities) ⁷⁸	Medium agreement, medium evidence (N=3) Outbreak: ^{31,97} No outbreak ²⁹	Low evidence (N = 1) ⁹⁸ investigate heavy rainfall, but defined 'flood' periods caveating that this did not necessarily correspond with actual flood waters

Alcayna et al. Climate-sensitive disease outbreaks in the aftermath of extreme climatic events: A scoping review. 2022. *OneEarth*.

Climate-Informed Early Warning Systems

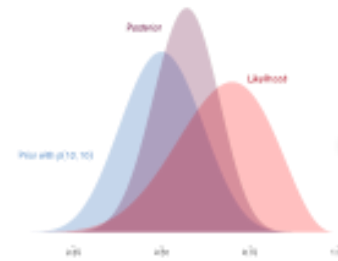
Climate Reanalyses



Land-based observation



Disease Surveillance



Statistical model



Seasonal climate forecast



Diseases forecast

Climate-Informed Early Warning Systems

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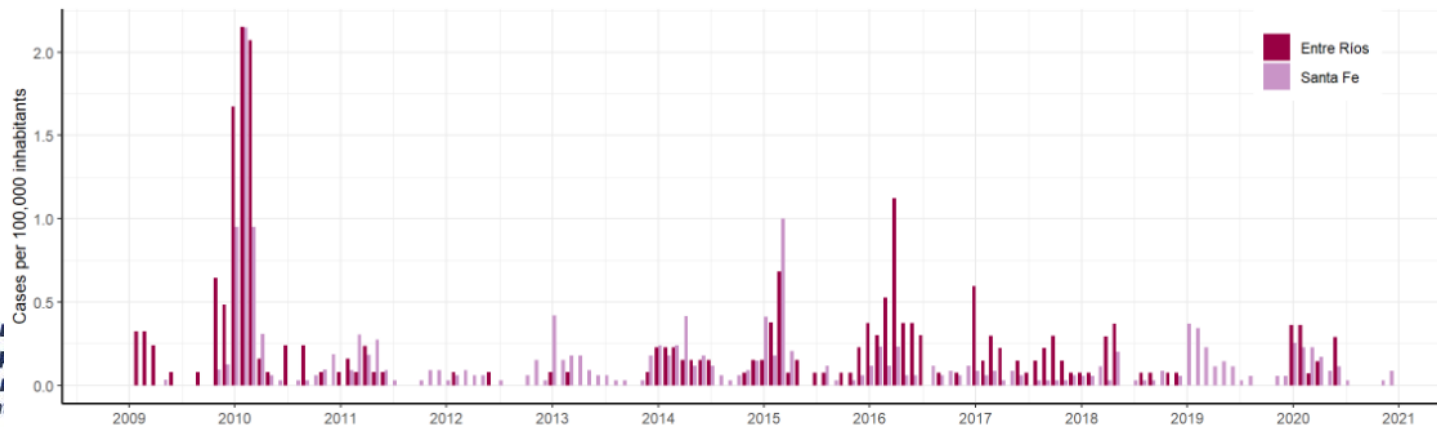
Tools Share

Research articles

Towards a leptospirosis early warning system in northeastern Argentina

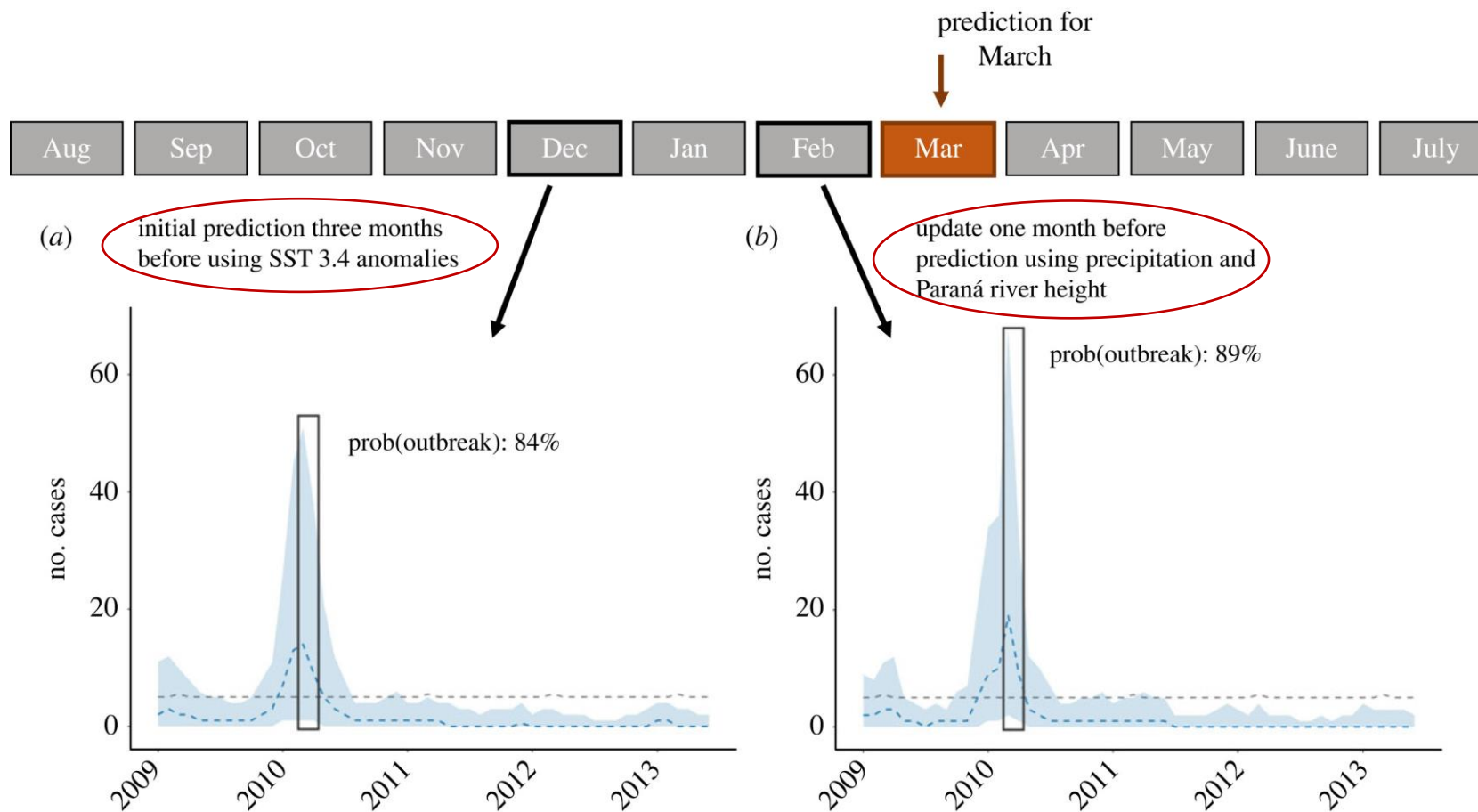
Martín Lotto Batista[†], Eleanor M. Rees[†], Andrea Gómez, Soledad López, Stefanie Castell, Adam J. Kucharski, Stéphane Ghazzi, Gabriela V. Müller and Rachel Lowe 

Published: 17 May 2023 | <https://doi.org/10.1098/rsif.2023.0069>



◆ Met. Station
▲ River height

Climate-Informed Early Warning Systems



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WORK IN PROGRESS



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Aim

To create user-friendly tools to quantify the combined impact of hydrometeorological extremes on disease risk and predict the probability of outbreaks using **observed and forecast hydrometeorological indicators.**

Case Studies



— Barbados



Brazil



South Sudan



Nepal



With support from



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Tool Overview

The IDExtremes tool will be based on an **R package** developed by the project.

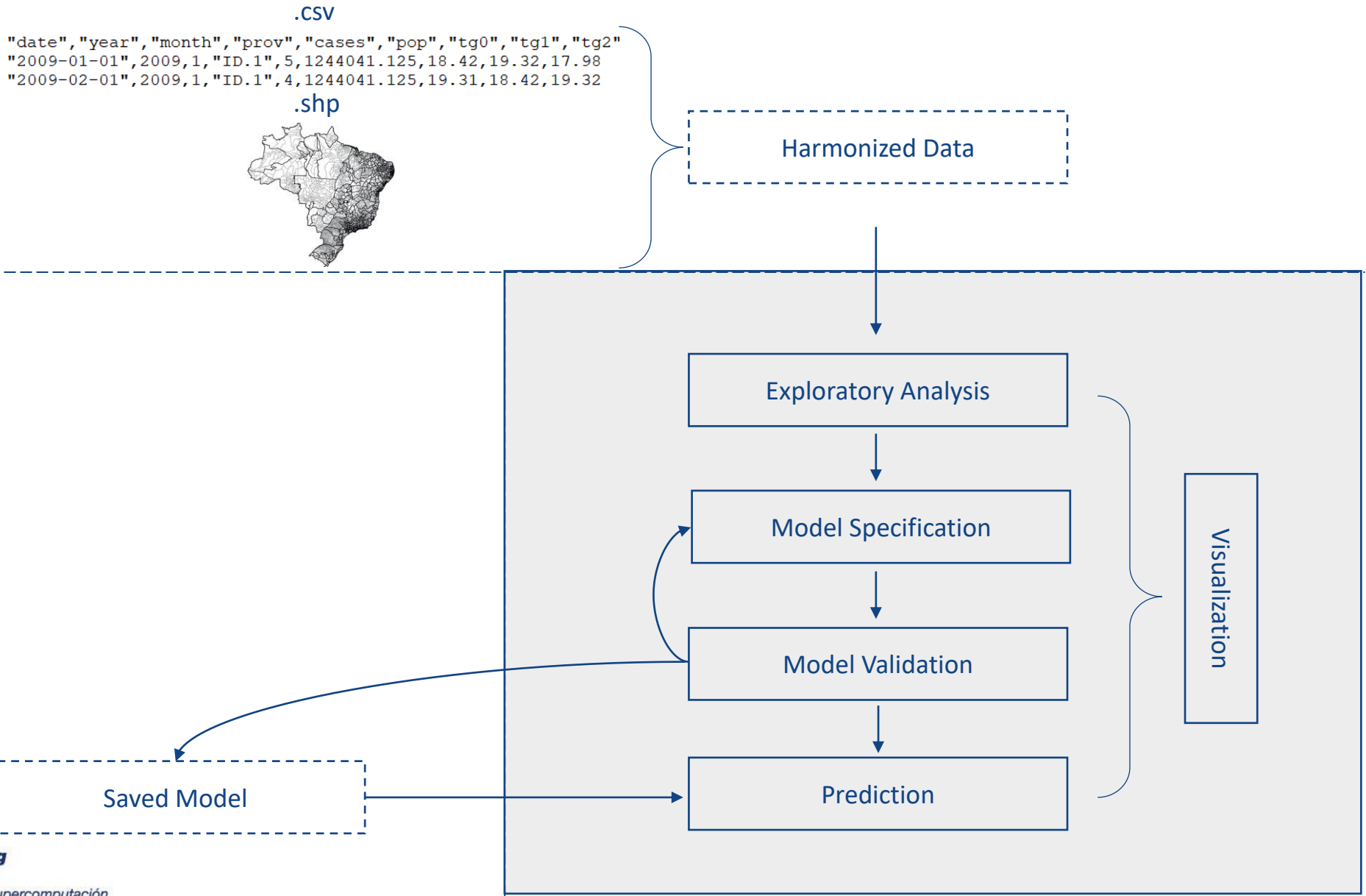
The R package should allow the users to produce probabilistic forecasts of climate sensitive infectious diseases including hydrometeorological indicators.

At the moment 4 steps are identified:

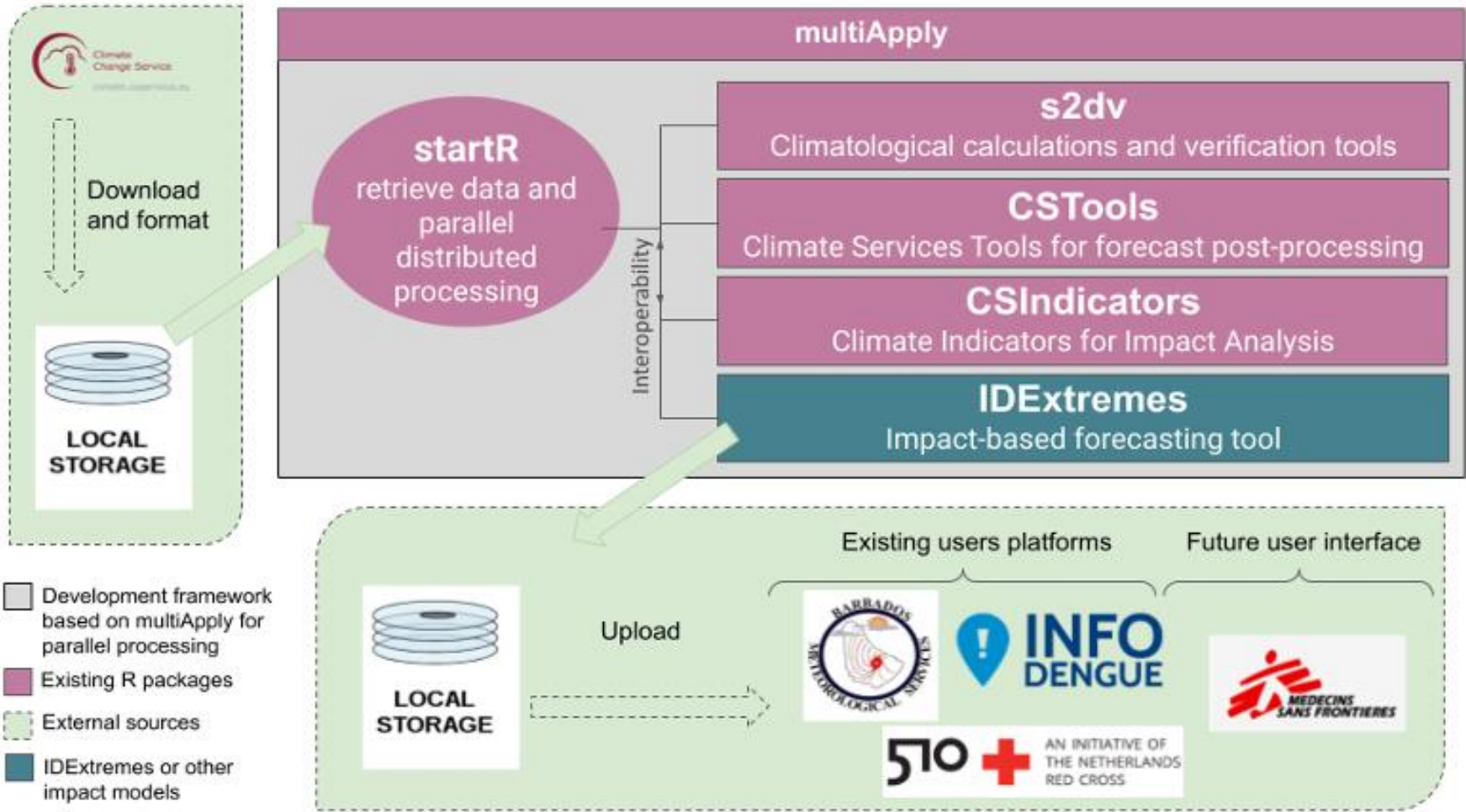
1. Exploratory analysis
2. Model specification
3. Model validation
4. Model prediction

All four steps should be accompanied by appropriate visualisation functions.

Tool Overview



Tool Overview

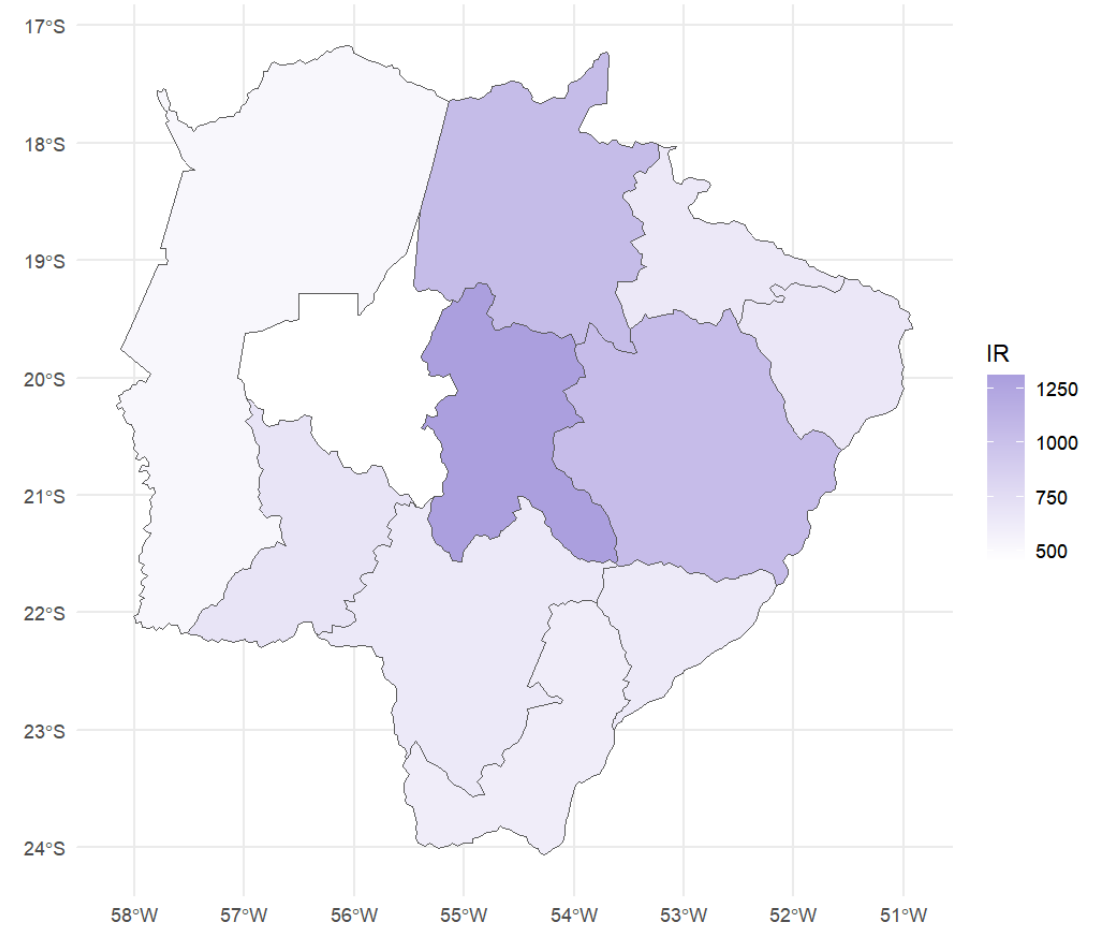


Harmonized Data



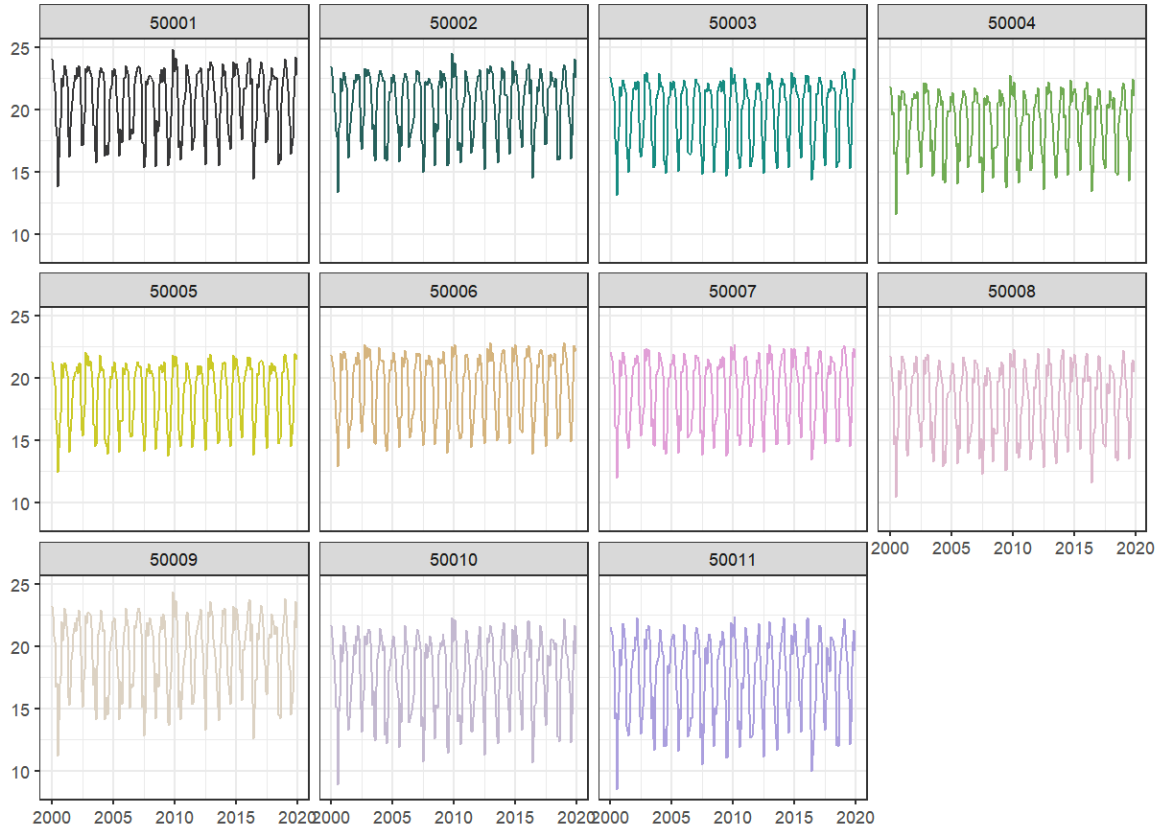
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"date", "year", "month", "prov", "cases", "pop", "tg0", "tg1", "tg2"  
"2009-01-01", 2009, 1, "ID.1", 5, 1244041.125, 18.42, 19.32, 17.98  
"2009-02-01", 2009, 1, "ID.1", 4, 1244041.125, 19.31, 18.42, 19.32
```

Incidence Rates (100'000 py) for Dengue in Mato Grosso do Sul



1. Exploratory Analysis

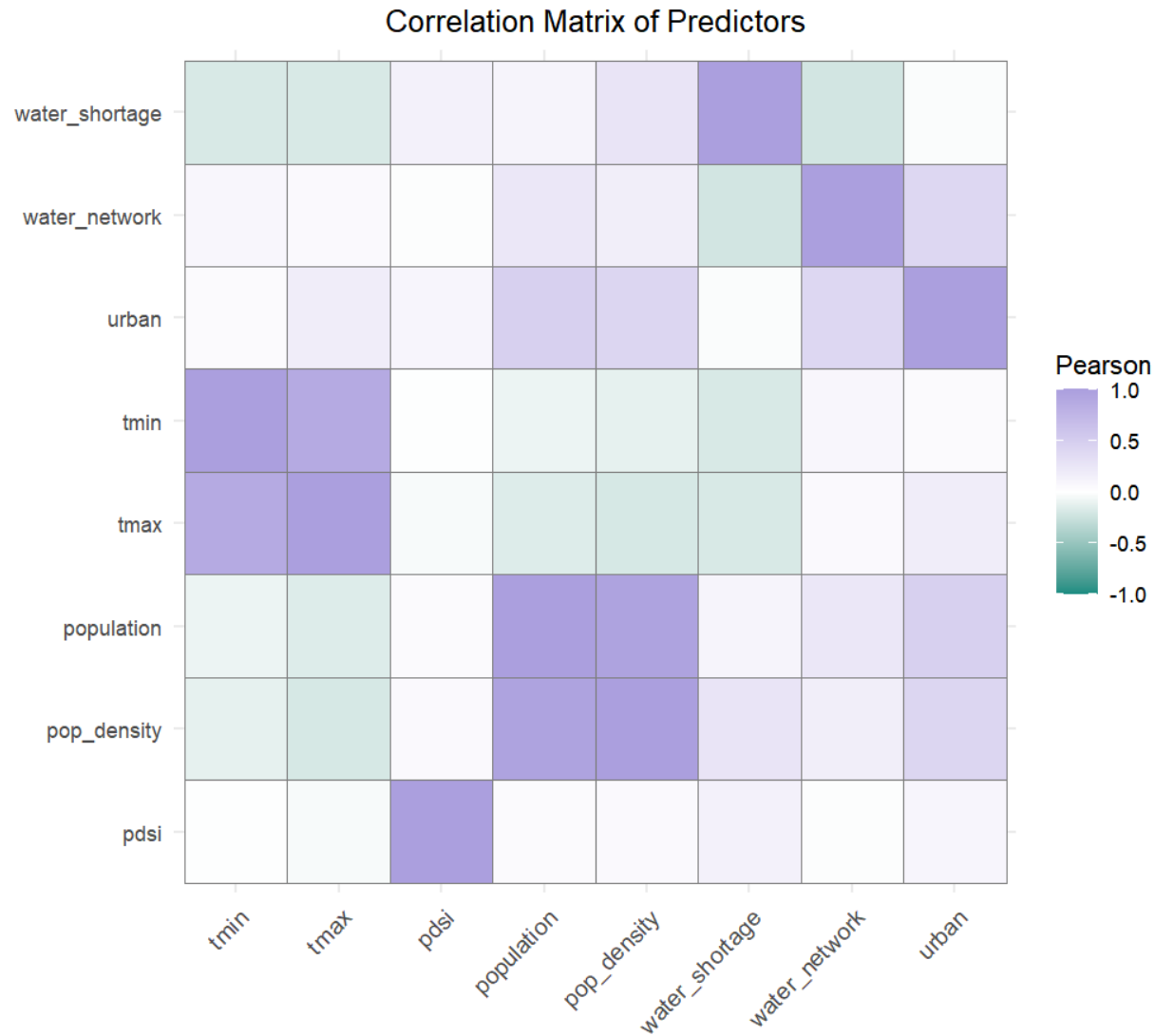
Minimum Temperature



Dengue Incidence Rates



1. Exploratory Analysis



2. Model Specification

Bayesian Spatio-temporal models

$$Y_{it} \sim \text{Poisson}(\mu_{it})$$

$$\ln(\mu_{it}) \sim 1 + \ln(E_{it}) + X_{it}^T \beta$$

Where for each area i and/or time unit t :

- μ_{it} is the expectation of the observed cases,
- E_{it} are the expected number of cases,
- X_{it}^T is a set of spatio-temporal hydrometeorological parameters of interest (lagged)

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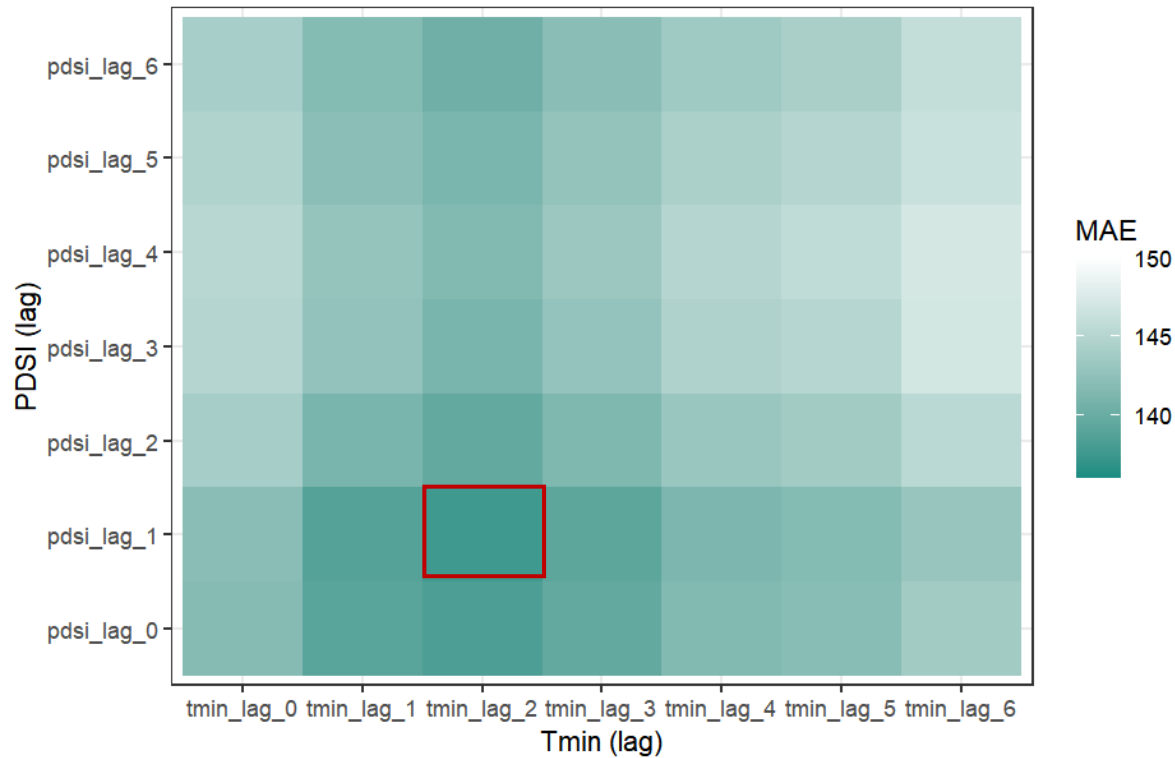
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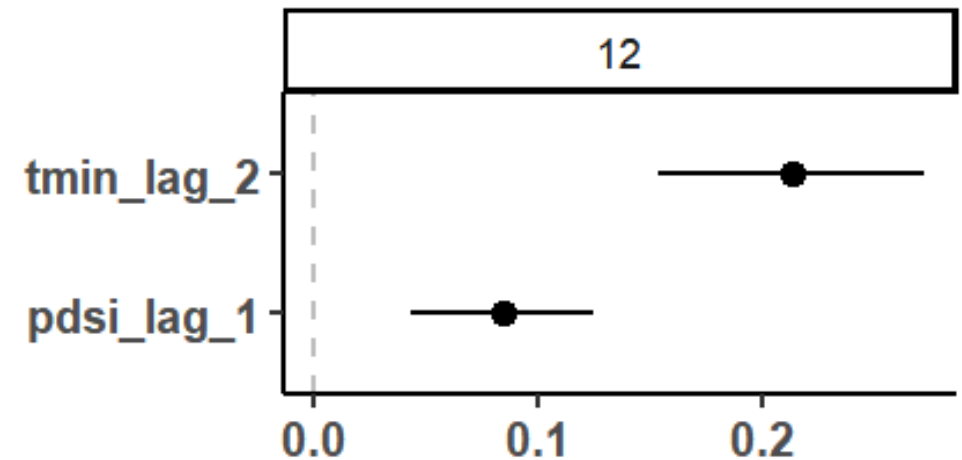
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- X_{it}^T is a set of spatio-temporal hydrometeorological parameters of interest (lagged)
- φ_i is the set of spatial random effects
- δ_t is the set of temporal random effect
- ξ_{it} is a set of random effects for space-time interaction

2. Model Specification (Outputs)

Allow the user to specify and evaluate performance of the fitted models



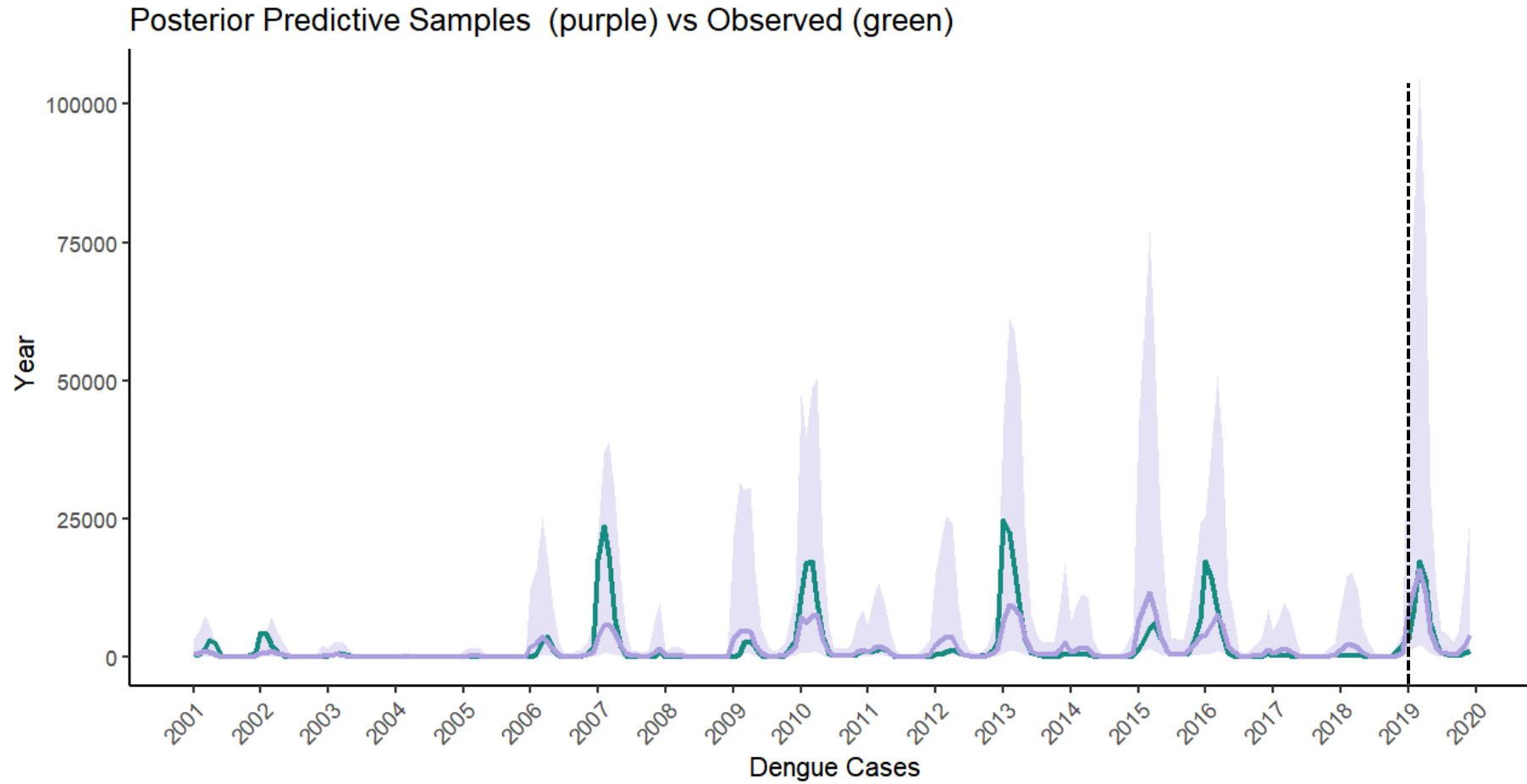
Check Model Performance



Evaluate Size and Timing of Associations

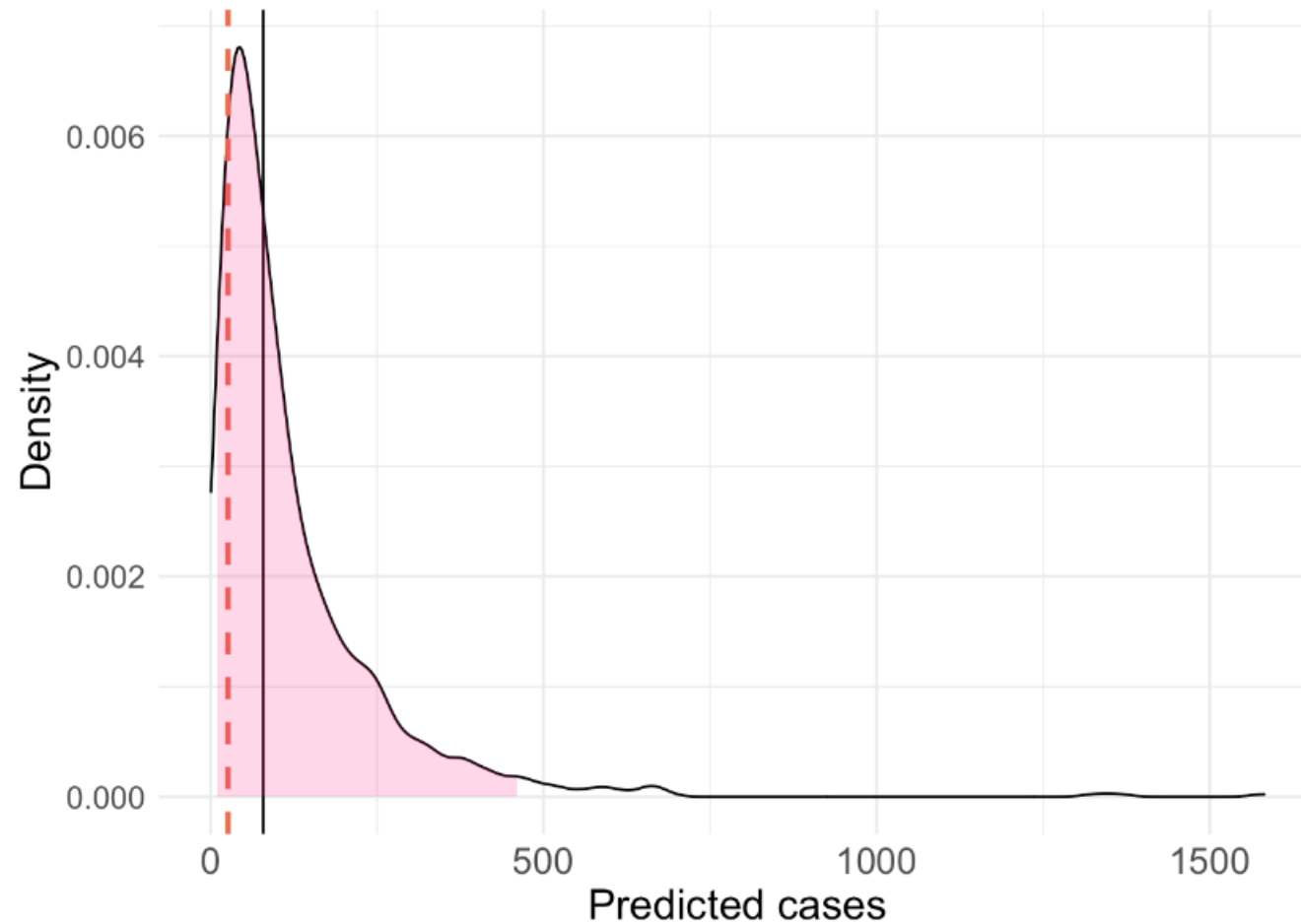
3. Model Validation

Allow the user to cross-validate the model with out-of-the-sample predictions



4. Model Prediction

Bayesian inference takes into account uncertainty propagation. In this context outbreaks probability can be estimated by computing the number of posterior predictive samples above a threshold level for each time and space unit.



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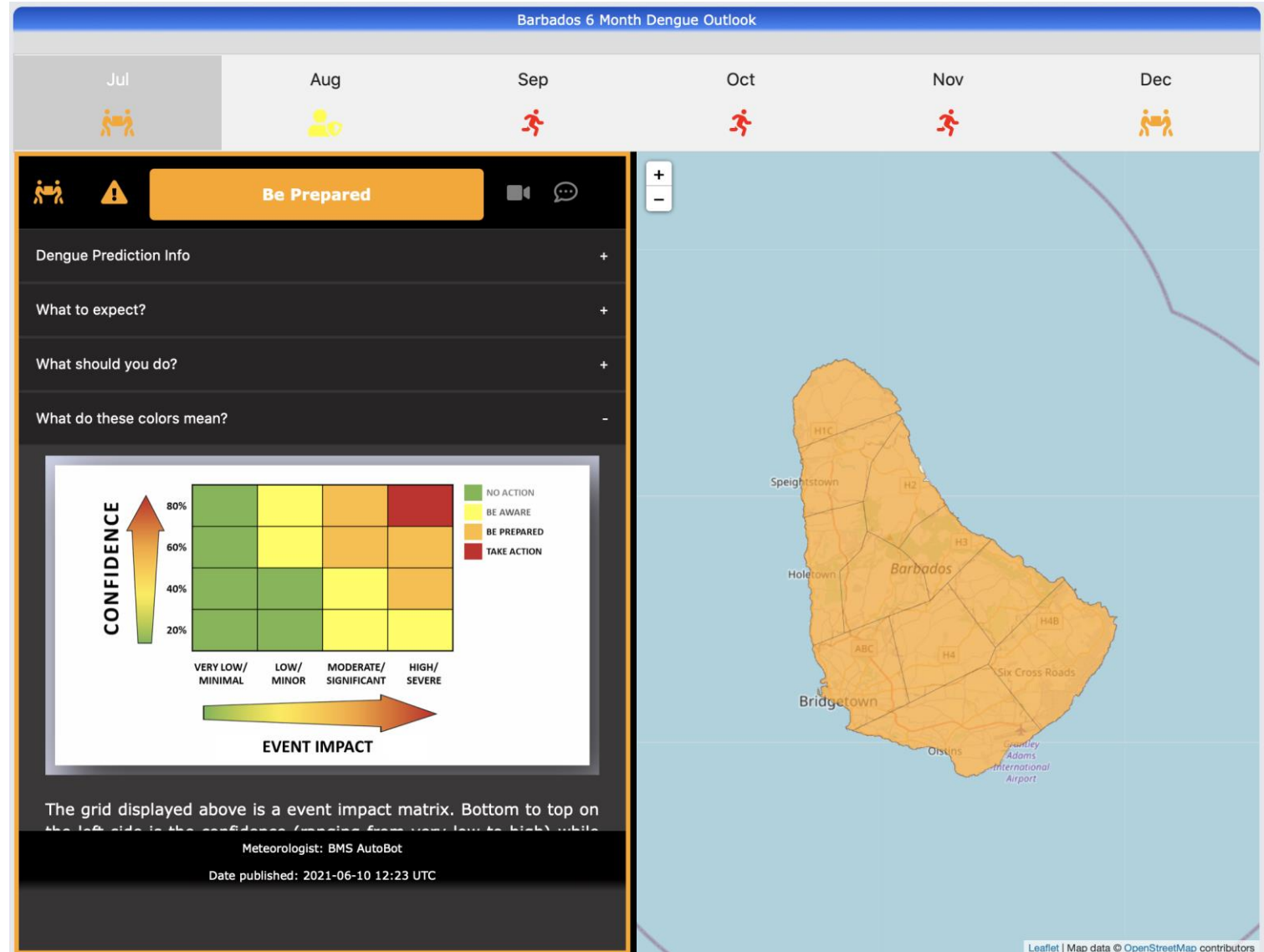


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Translating results into relevant indicators for Public Health

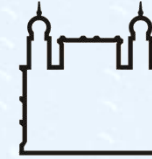
Prototype developed by Barbados Meteorological Services



IDExtremes consortium



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Fundação Oswaldo Cruz



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AN INITIATIVE OF
THE NETHERLANDS
RED CROSS



**INFO
DENGUE**

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Raquel Lana



Bruno Carvalho



Kim van Daalen



Giovenale Moirano



Chloe Fletcher



Alba Llabrés



Prof. Rachel Lowe



Martin L. Batista



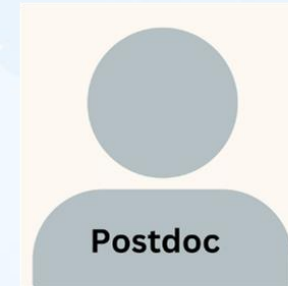
Daniela Lurhsen



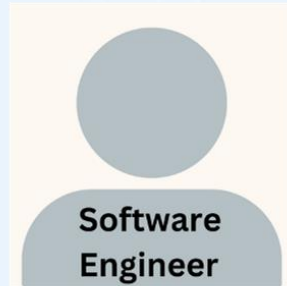
Remy Hoek Spaans



Sophie Belman



Postdoc



Software Engineer



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Obrigado

Perguntas?

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