



Forecasting Outbreaks (Part 3)

Prof. Dr. Axel Kroeger

Albert-Ludwigs-Universität Freiburg - Centre
for Medicine and Society, Germany

Ass. Prof. Dr. Laith Hussain

University of Gothenburg, Sweden

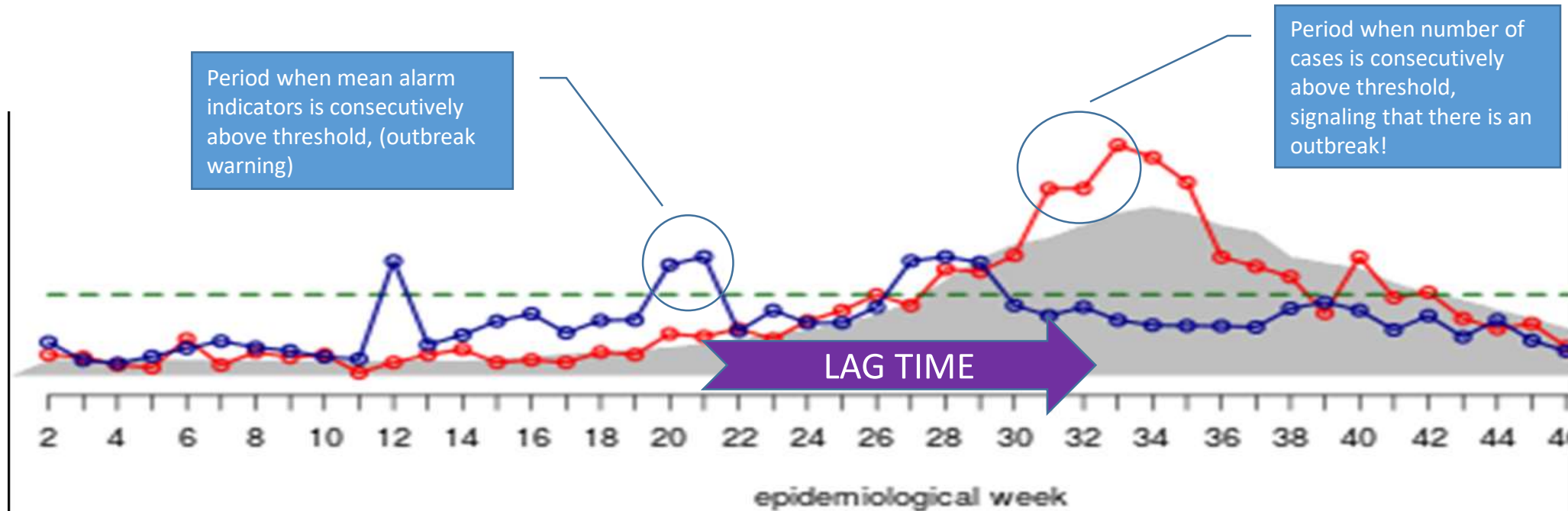
Supported by TDR and ECH WHO,
Geneva



Supported by TDR and ECH WHO, Geneva; Freiburg University, Germany; Gothenburg University, Sweden

Definitions (3): Lag time

- The *Lag time* or *prediction period* is the number of weeks from alarm signal to start of the outbreak



The lag time depends particularly on climate variables:

- **Short lag time** = favourable meteorological conditions for vectors (higher temperature and humidity, rainfall)
- **Long lag time** = Unfavourable conditions for the vector e.g. too hot, too cold, too much rain, too dry

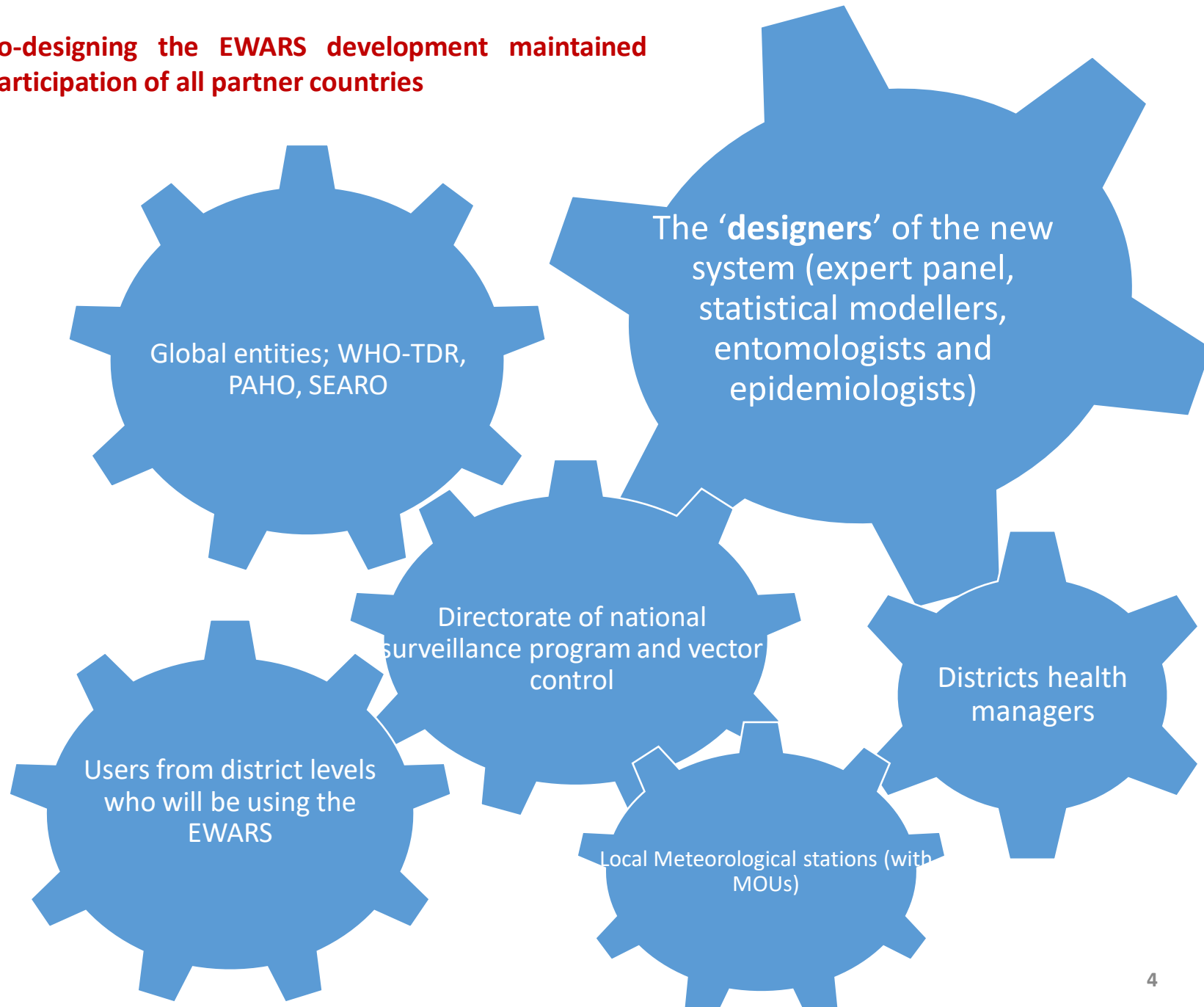
The EARLY WARNING AND RESPONSE SYSTEM (EWARS plus, TDR/WHO)

- The development and validation
- Literature reviews
- How is EWARS used by countries?
- How has it been tested so far

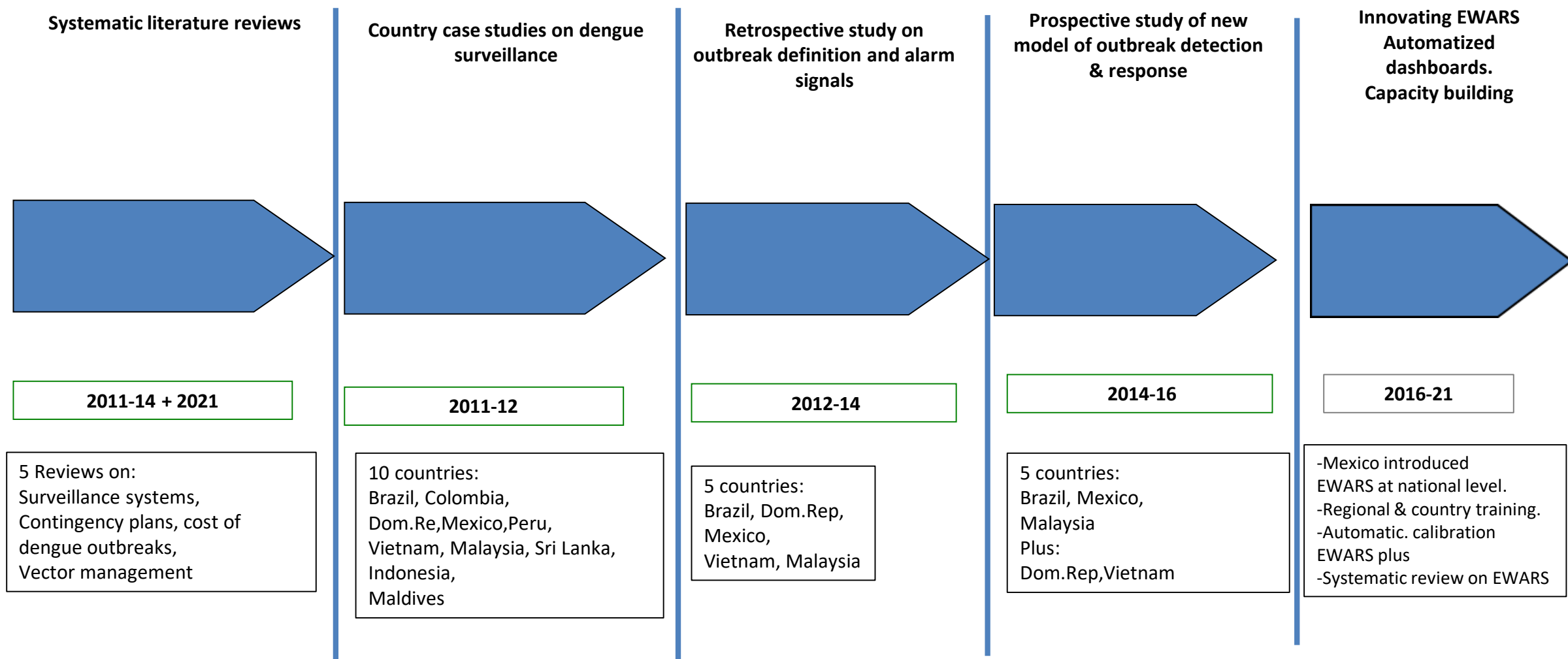
Co-designing the EWARS development maintained participation of all partner countries

Co-design approach

“Co-design involves and maintains working together to design a new product, making full use of each other’s knowledge, resources and contributions, to achieve better outcomes or improved efficiency”



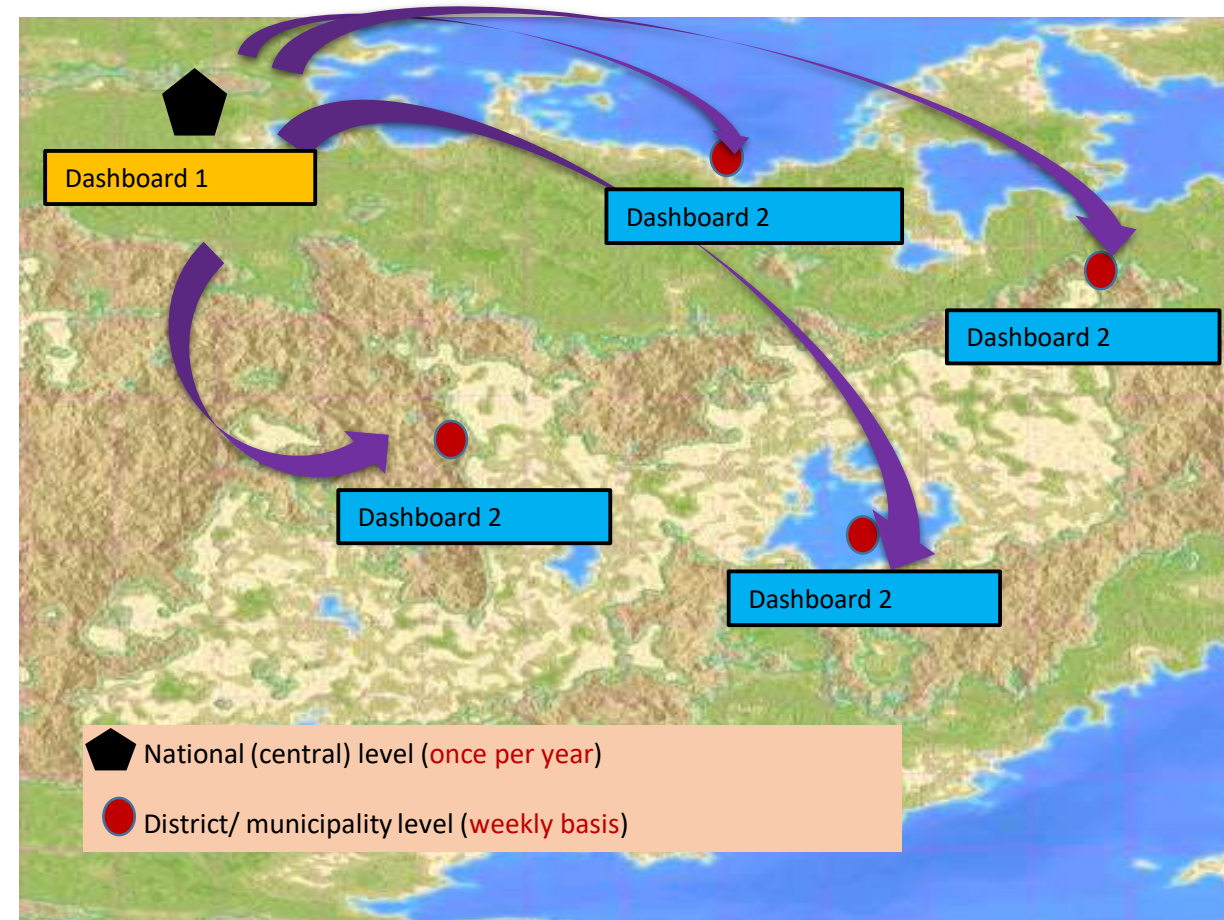
Brief history of EWARS development



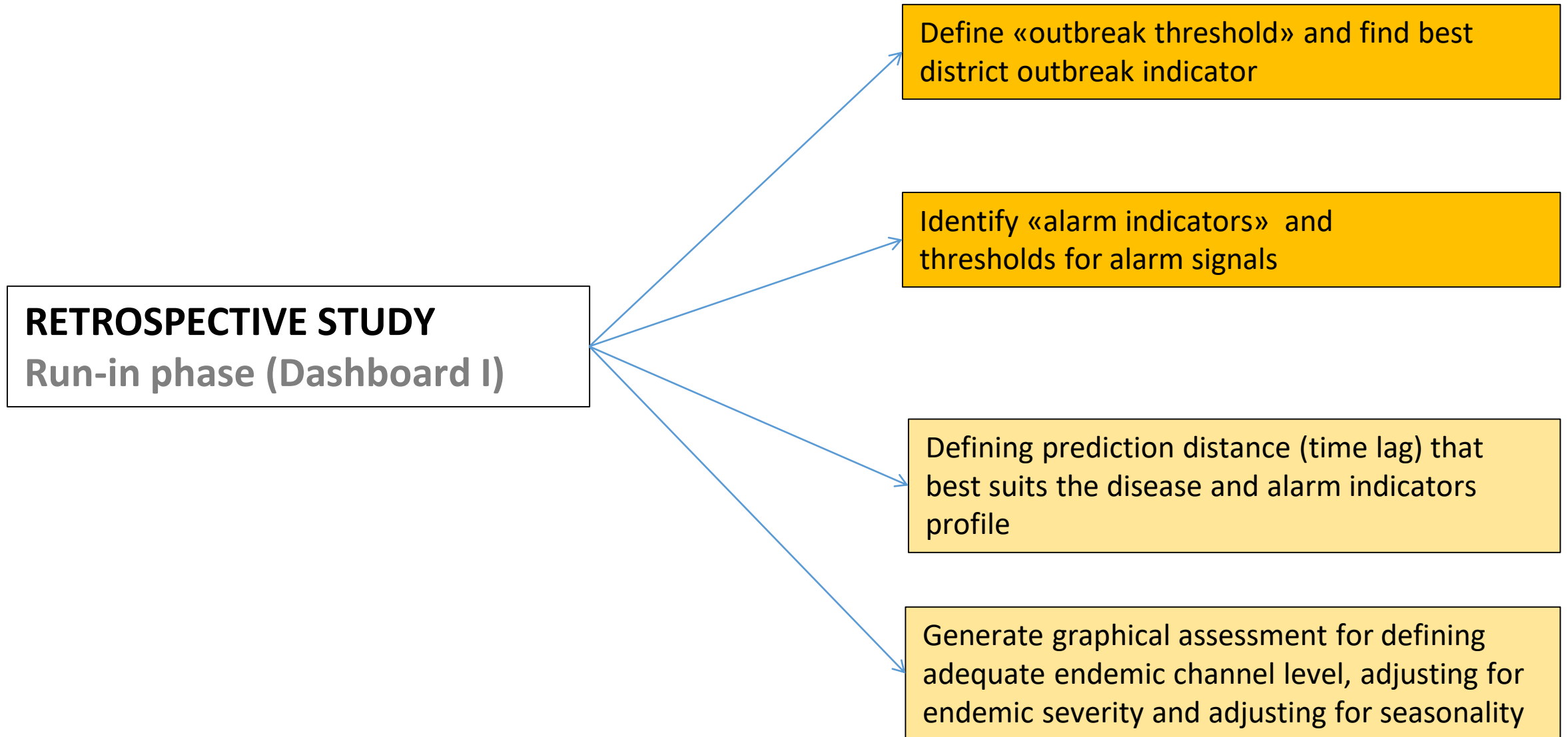
CHARACTERISTICS OF EWARS TDR/WHO (1)

ONE HUB: https://alramadona.shinyapps.io/Ewars_dashboard/

- **Officers at national level (Dashboard I)**
 - ❖ *1st dashboard*: validate model ++ oversee the outbreak prediction at district level ++ train sub-national level
- *Parameters are automatically and instantly linked to sub-national level via web*
- ❑ **Officers at district/ municipality level**
 - ❖ *2nd dashboard*: weekly prospective data input, interpretation & action
- ❑ *This platform facilitates a free sustainable DATABASE for the surveillance data*



CHARACTERISTICS(2): Calibrating the model using first half of data



CHARACTERISTICS(3): Evaluating the model using second half of data

Determining validity of prediction (Dashboard I, evaluation phase)

Sensitivity (of the alarm)

The proportion of alarms that successfully predict defined outbreaks

i.e. no. of correct outbreak alarms/ total outbreaks

PPV

(Positive Predictive Value)

The proportion of true alarms out of all alarms
i.e. no. of correct alarm periods/ total no. of alarm periods

Examples:

Sensitivity = 90%

9 out of 10 outbreaks have been correctly detected

PPV= 70%

7 out of ten alarm signals were correct

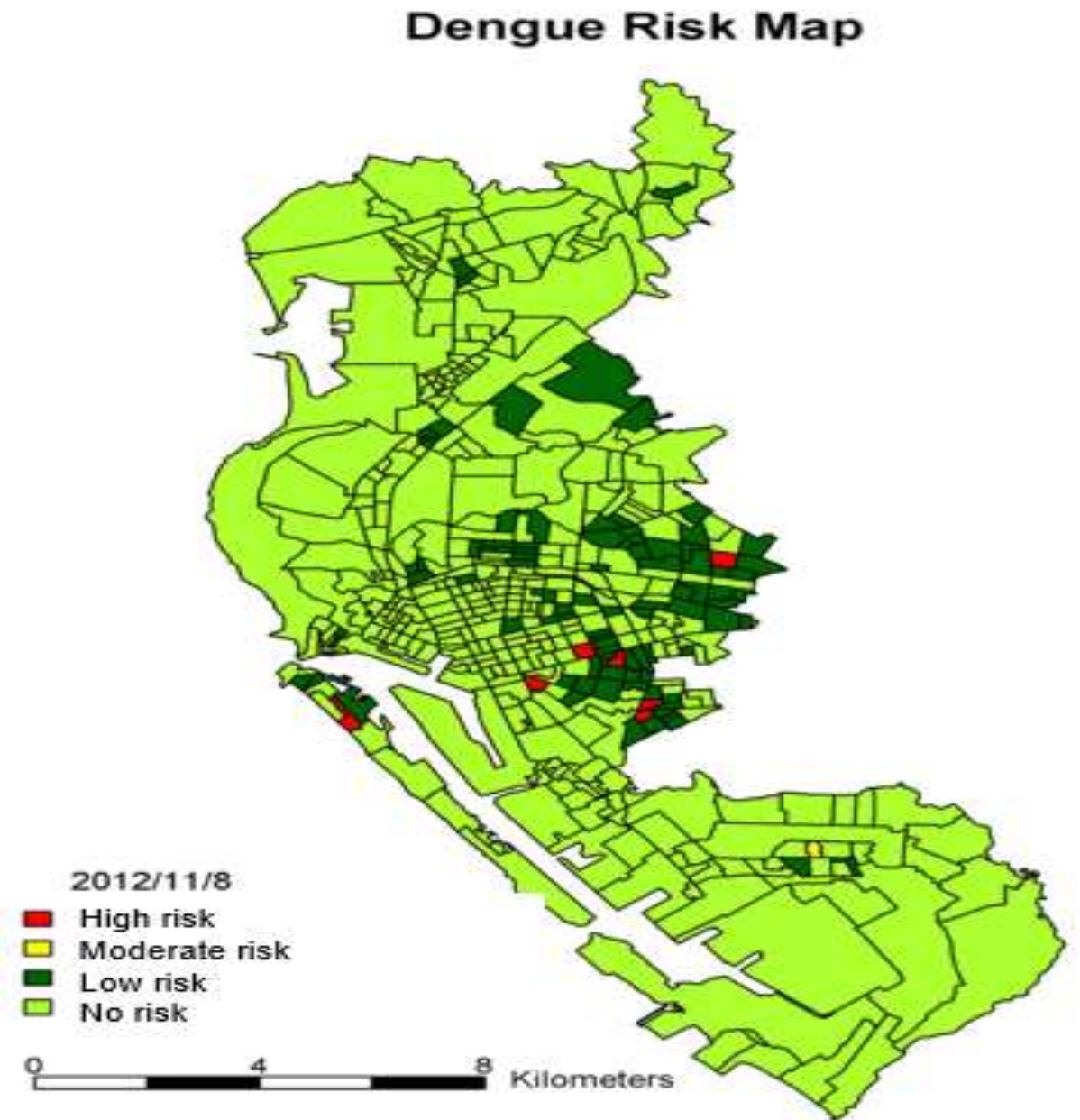
EXAMPLE for validity testing of outbreak alarms: PPV and Sensitivity (Mexico)

Alarm Indicator	Outbreak indicator	Positive Predictive Value (%)	Sensitivity (%)
mean temp	hospitalized cases	72	81
rainfall	hospitalized cases	65	87
mean age	hospitalized cases	74	89
probable cases	hospitalized cases	83	100
Ovitrap (%positive traps)	hospitalized cases	60	79
humidity	hospitalized cases	50	94
Serotype	hospitalized cases	75	100
multiple indicators*	hospitalized cases	77	84

* temperature, rainfall, mean age, probable cases, positive ovitrap & humidity

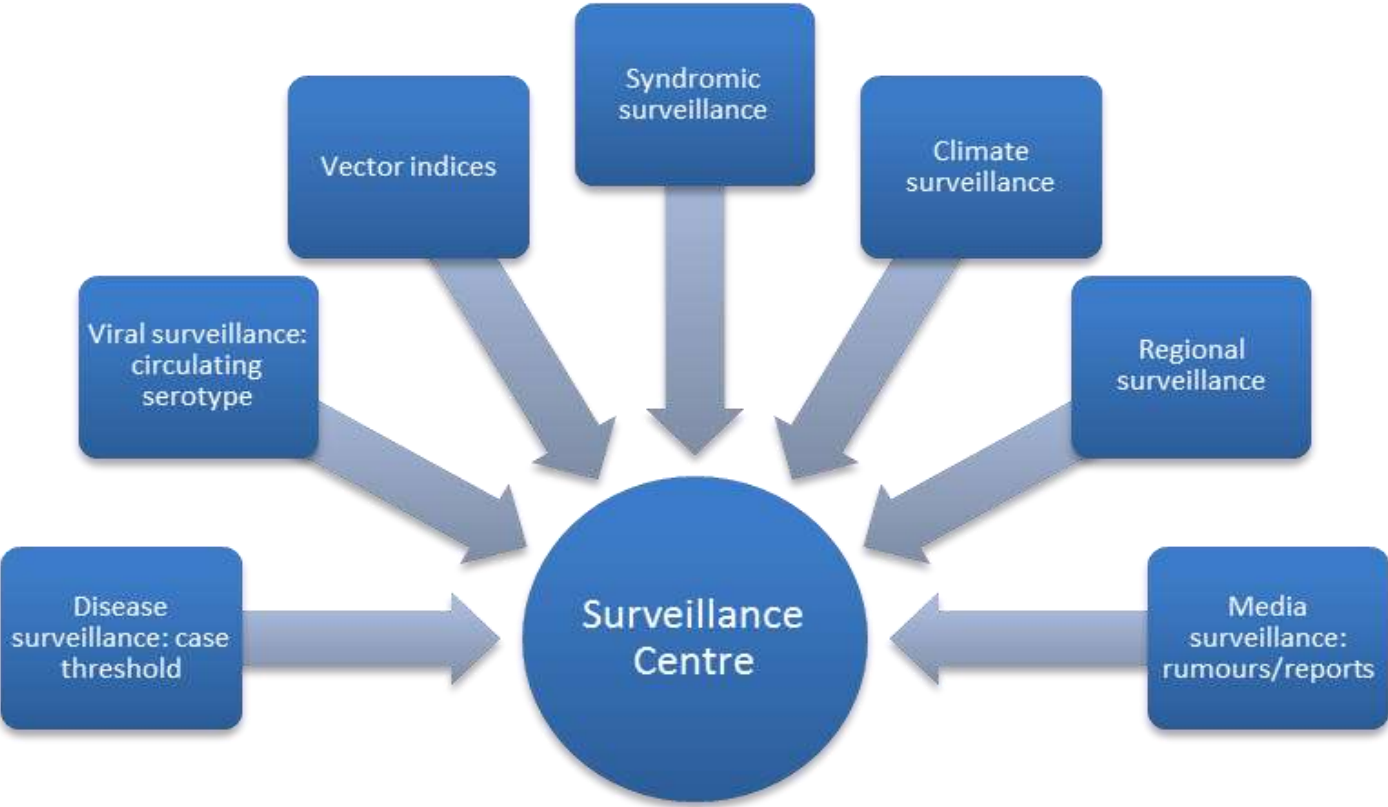
CHARACTERISTICS(4): Spatial analysis

- ❑ Provides a rapid visual summary of spatial information
- ❑ Crucial for describing the spatial and temporal variation of the disease
- ❑ Identifies areas of unusually high risk areas (hot spots)



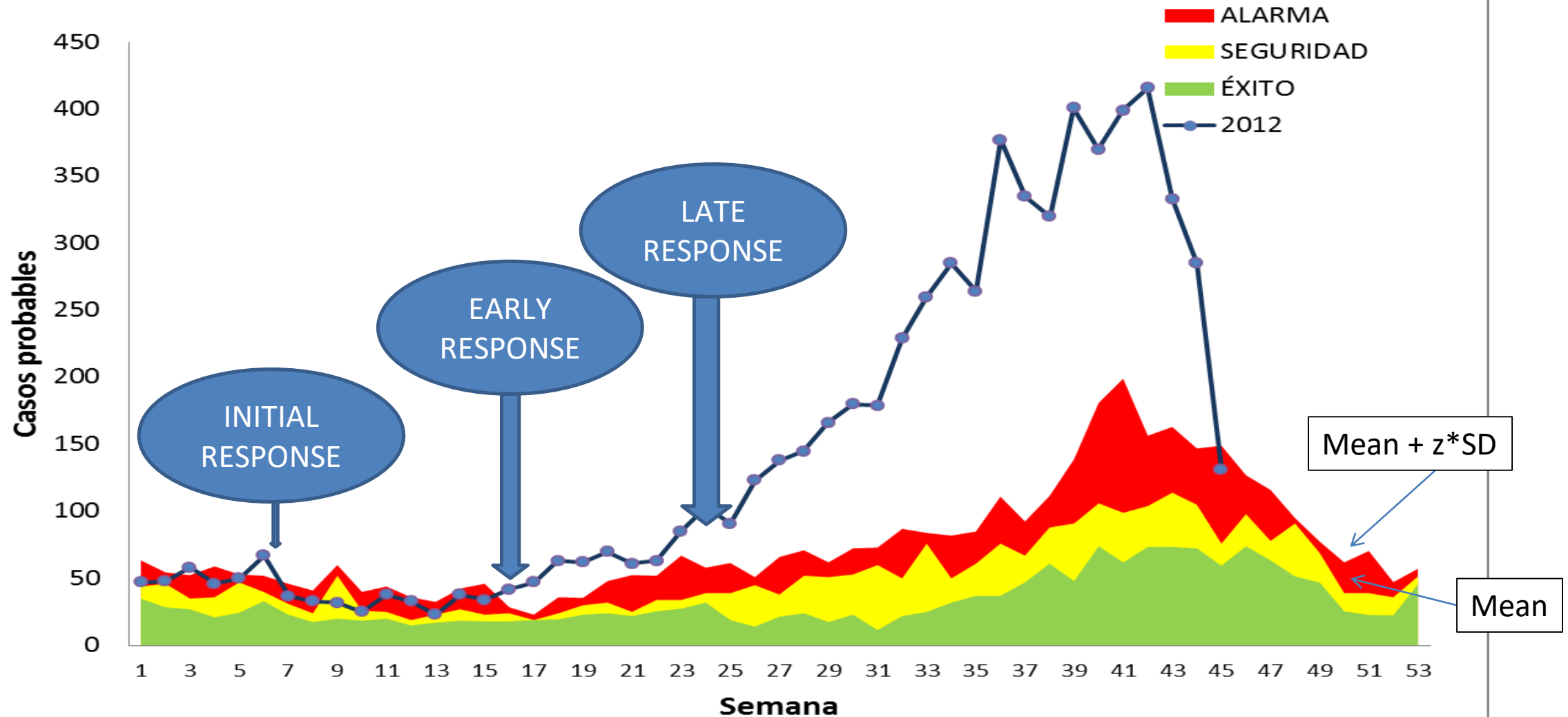
Source: Ciaran Nugent, Forest Service, Department of Agriculture, Food and the Marine (DAFM).

CHARACTERISTICS(5): Level of alarm leading to staged response



Initial response	Early response	Late response
Outbreak Investigation	Activate communication channels	Full implementation of contingency plan including M&E activities and stopping rules
Assure contingency plan is available	Risk communication and outbreak declaration	
Update background information	Consider M&E activities	
Enhance surveillance	Implement community based risk reduction activities	
Enhance routine activities	Inform clinicians and prepare dengue treatment areas	
Complete preparatory activities including training	Ensure guidelines for case detection, treatment and triage	
Prioritize risk areas	Intensify vector control activities	
	Communicate end of alert and response activities	

Staged response to an outbreak



BACKGROUND READING



Evidence-based tool

PLOS ONE

Early warning and response system (EWARS) for dengue outbreaks: Recent advancements towards widespread applications in critical settings

Laith Hussain-Alkhateeb , Axel Kroeger, Piero Olliaro, Joacim Rocklöv, Maquins Odhiambo Sewe, Gustavo Tejada, David Benitez, Balvinder Gill, S. Lokman Hakim, Roberta Gomes Carvalho, Leigh Bowman, Max Petzold

<https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0196811>

PLOS NEGLECTED TROPICAL DISEASES

REVIEW

Early warning systems (EWSs) for chikungunya, dengue, malaria, yellow fever, and Zika outbreaks: What is the evidence? A scoping review

Laith Hussain-Alkhateeb ^{1*}, Tatiana Rivera Ramírez ², Axel Kroeger², Ernesto Gozzer ³, Silvia Runge-Ranzinger ^{2,4}

<https://journals.plos.org/plosntds/article?id=10.1371/journal.pntd.0009686>

Supported by the



Federal Ministry
of Health



Policy Brief

**Innovation and Collaboration:
the EWARS Framework for
infectious diseases**

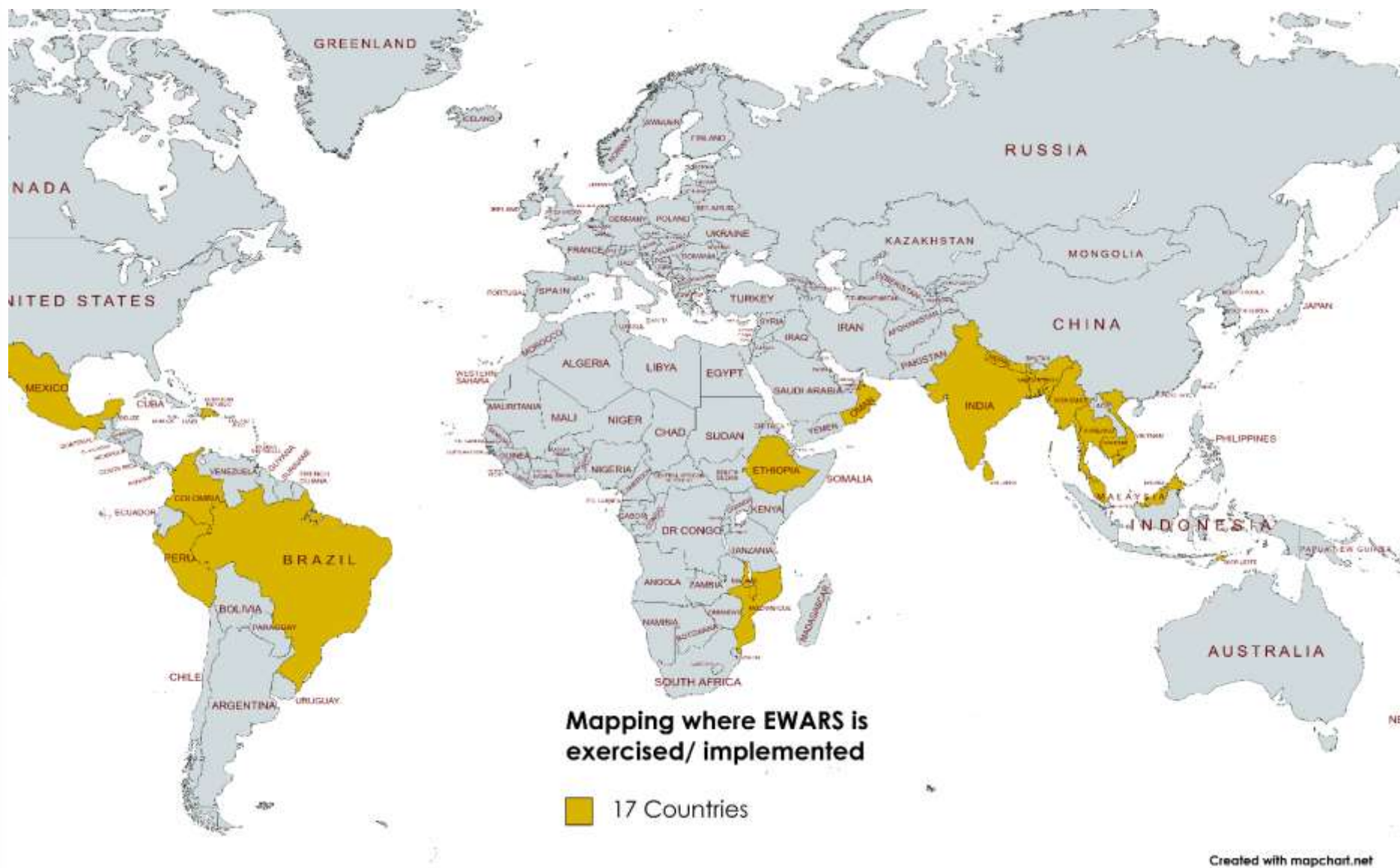
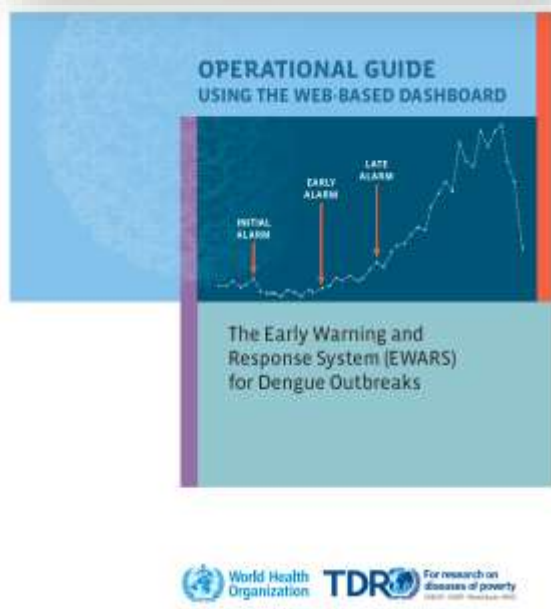
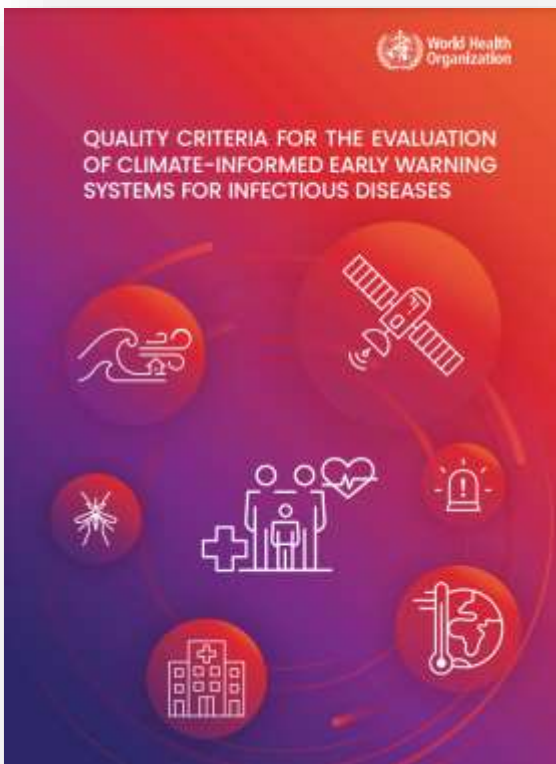


RESEARCH ARTICLE

Alarm Variables for Dengue Outbreaks: A Multi-Centre Study in Asia and Latin America

Leigh R. Bowman^{1,5*}, Gustavo S. Tejada², Giovanini E. Coelho³, Lokman H. Sulaiman⁴, Balvinder S. Gill⁴, Philip J. McCall¹, Piero L. Olliaro⁵, Silvia R. Ranzinger^{5,6}, Luong C. Quang⁷, Ronald S. Ramm⁸, Axel Kroeger^{1,5}, Max G. Petzold^{5,9}

<https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0157971>



<https://www.who.int/publications/i/item/10665345530>

<https://apps.who.int/iris/handle/10665/332323>



MANY THANKS FOR YOUR INTEREST