

ISI regional statistics conference

GHR tools

Streamlining climate and health workflows including data harmonization,
modelling and early warning systems

Carles Milà, Research Engineer

BSC Earth Sciences - Global Health Resilience

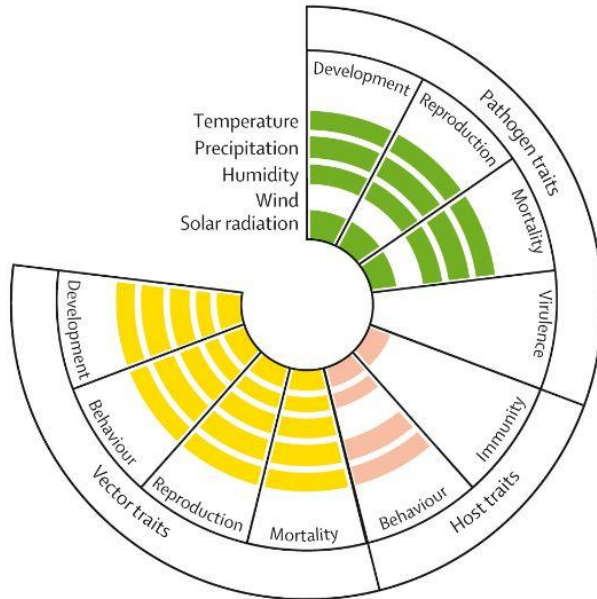
carles.milagarcia@bsc.es



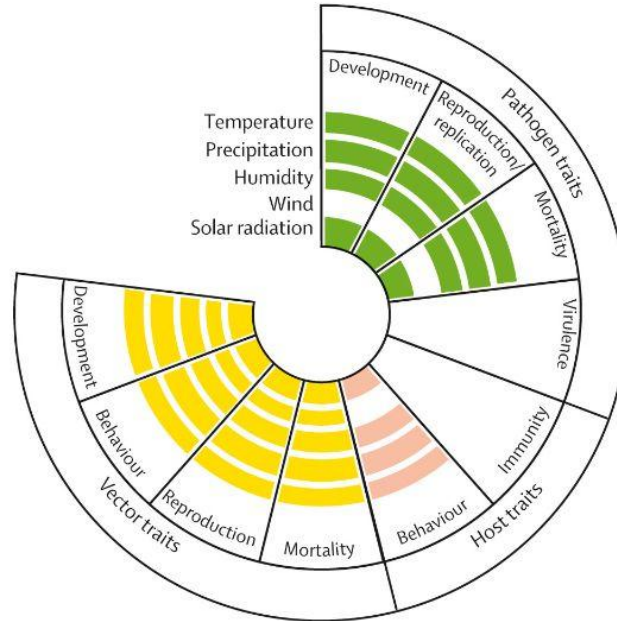
Climate variables associated with infectious diseases

- Evidence exists for the climate–pathogen trait association
- Evidence exists for the climate–host trait association
- Evidence exists for the climate–vector trait association
- The trait is not relevant for the disease
- Limited, insufficient, or unavailable evidence

Malaria



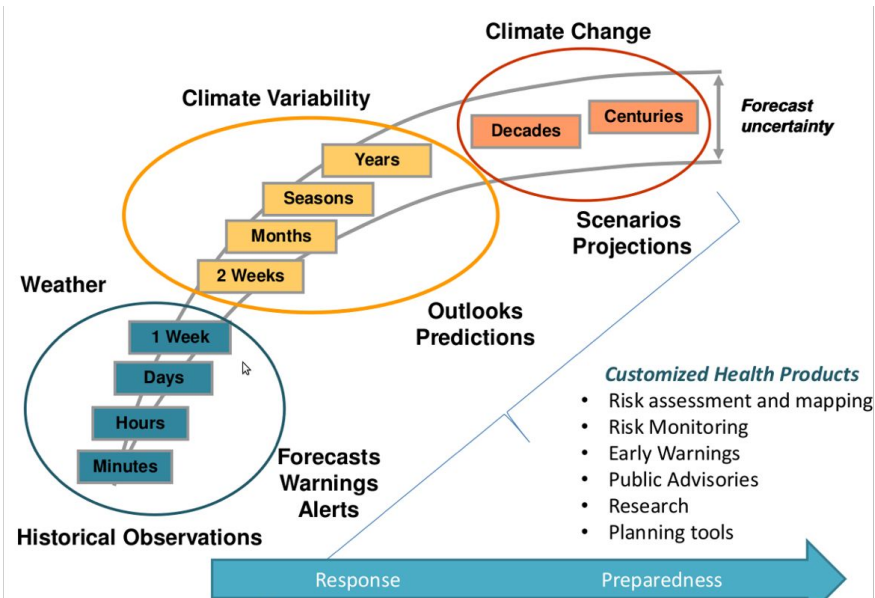
Dengue fever



Alcayna *et al.*
Lancet Planetary Health 2025

Early warning systems for climate-sensitive diseases

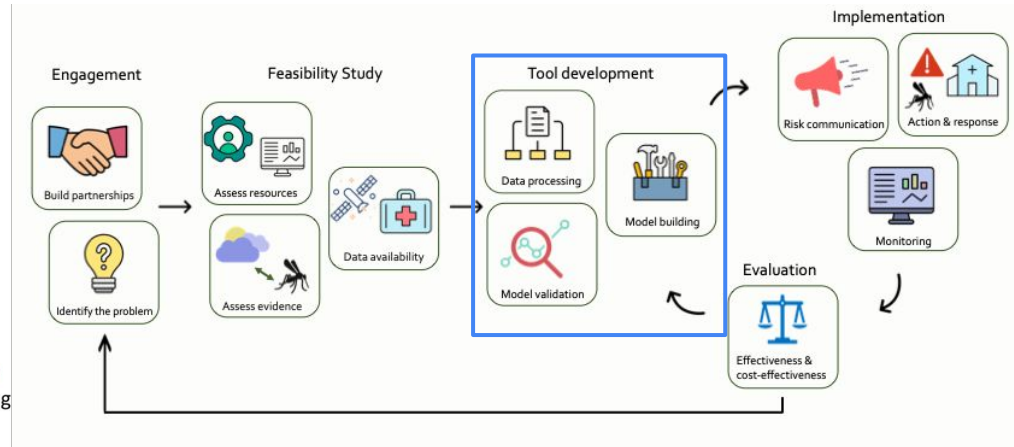
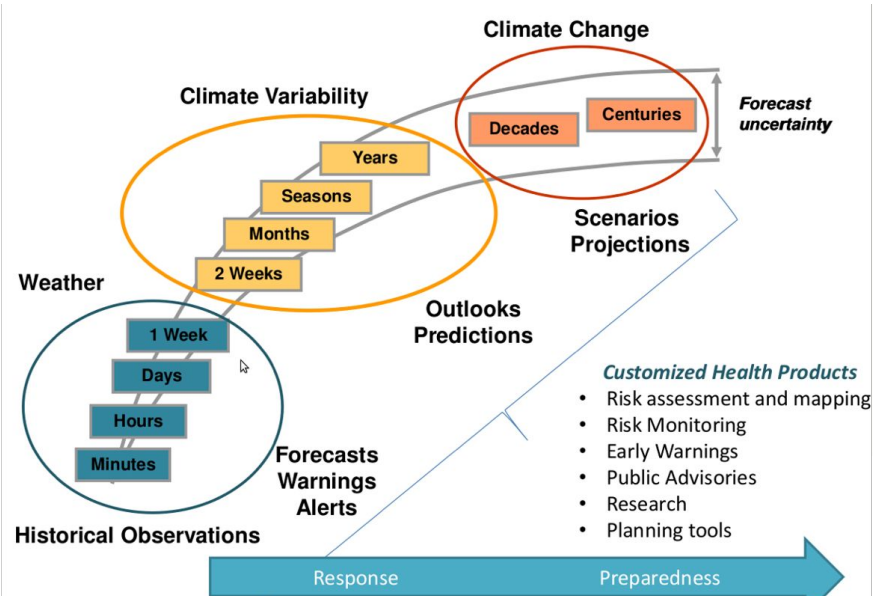
Climate-informed early warning systems integrate climate data to predict the risk of an infectious disease outbreak.



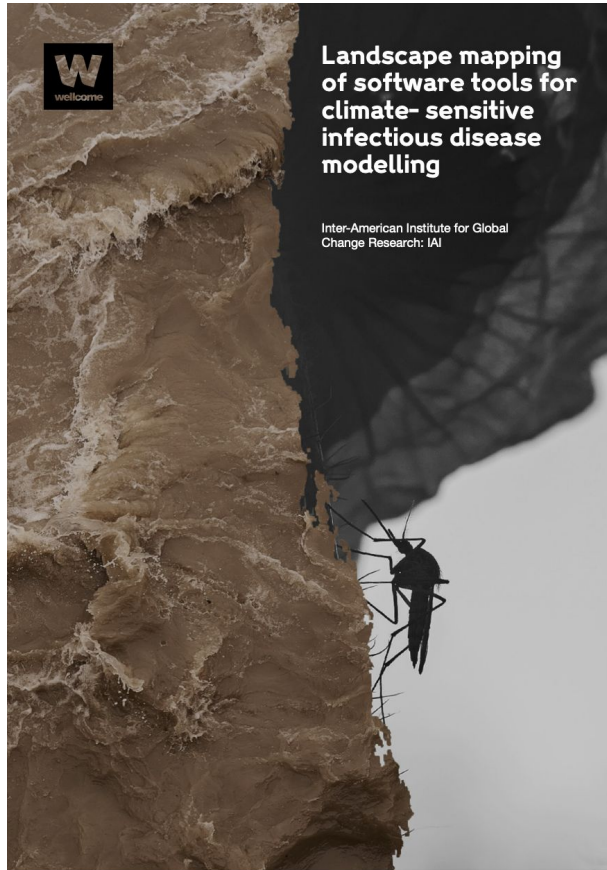
Courtesy: J. Shumake-Guillemot, WMO

Early warning systems for climate-sensitive diseases

Climate-informed early warning systems integrate climate data to predict the risk of an infectious disease outbreak.



Finch et al 2023



- A review of > 9,000 papers found that only 37 of them contained operational early warning systems.
- Lack of user-friendly, open and reproducible software to sustainably implement modelling tools anywhere in the globe.
- Need to develop open-source and free software for climate-sensitive early warning systems.

About Global Health Resilience (GHR)



Harmonizing multi-scale spatiotemporal data for health in
climate change hotspots

Disease surveillance data

Demographic and
socio-economic indicators

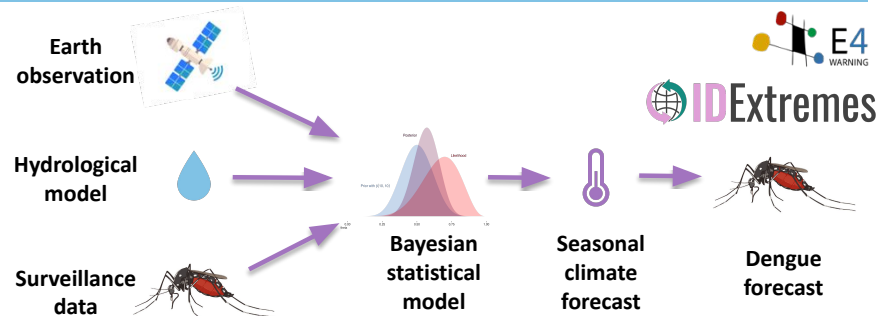
Land use and cover data

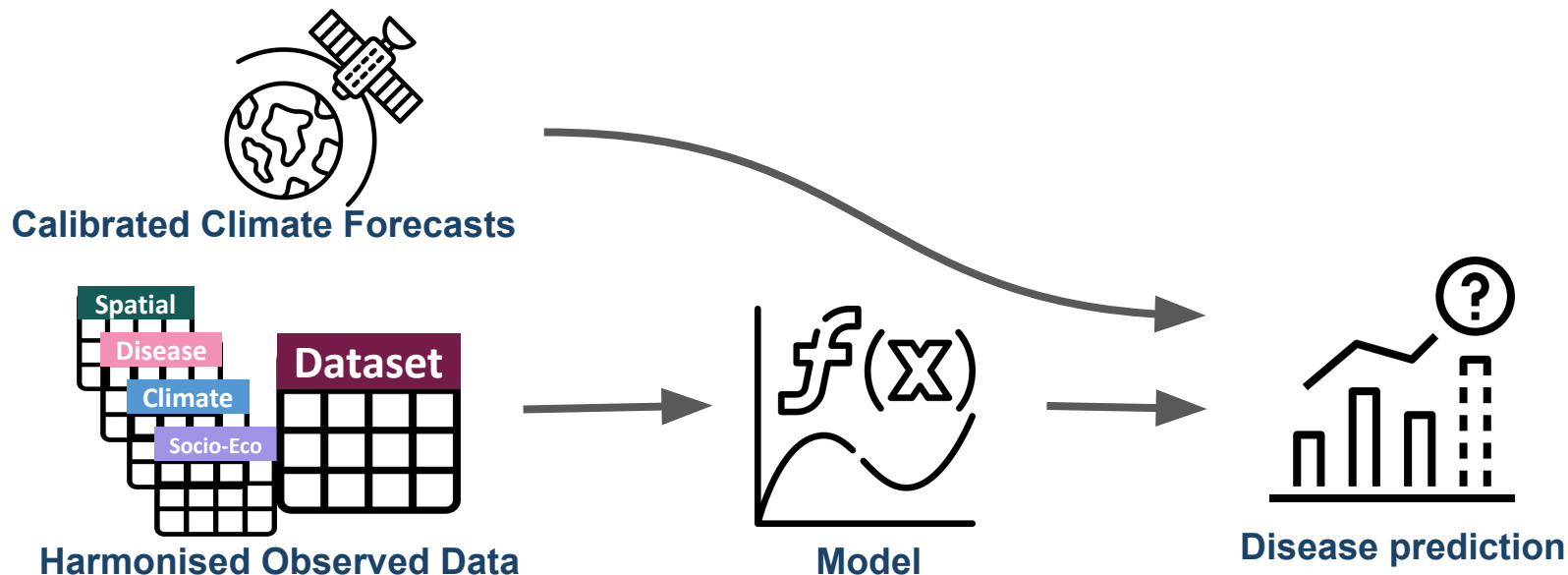
Weather & climate data

Harmonized space-time
data for small areas

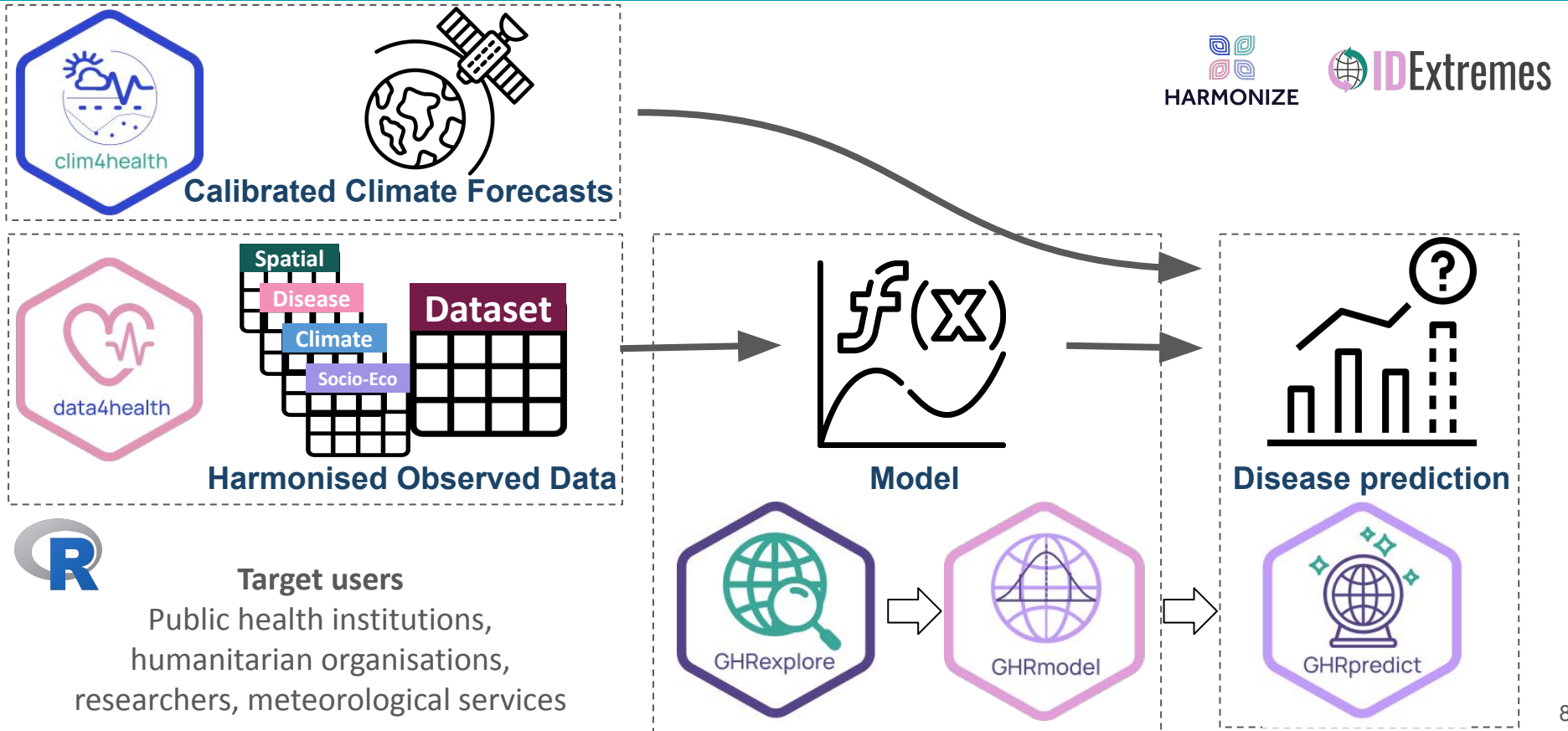


Developing early warning systems for climate-sensitive
infectious diseases





How GHRtools fit into the disease risk modelling and forecasting framework





An R package to facilitate the access, preprocessing, and aggregation of health data at customized spatiotemporal resolutions.



DANIELA LÜHRSEN
DATA SCIENTIST



DR RAQUEL LANA
RESEARCHER



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Expected release date:
July 2026

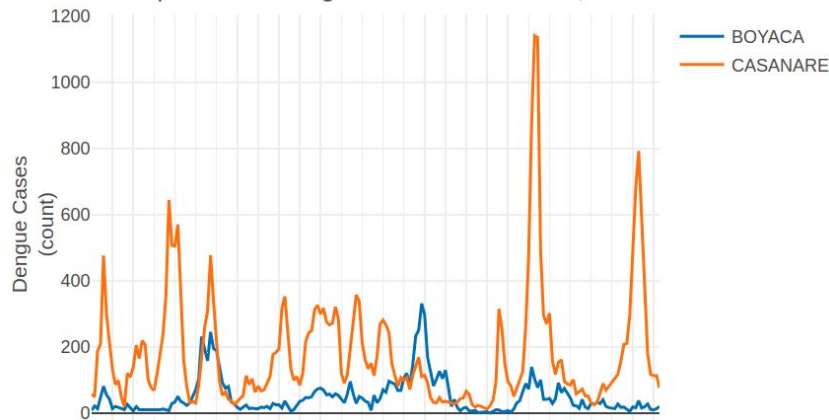
Unprocessed health data

	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X
1	DT_CLOS	DT_BIRTH	DT_NOTI	UF_NOTI	MUNI_NOTI	AGE	SEX	PREG	RACE	UF_INF	COPASINF	MUNI_INF	CLASS	CRIT	
2	2015-03-05	1979-07-27	2015-03-02	12	120020	35F	6NA	12	1	120033	10	2			
3	2015-02-02	2000-09-22	2015-01-22	12	120020	14F	6NA	12	1	120033	10	2			
4	2015-06-17	2005-05-10	2015-05-17	12	120020	10F	6NA	12	1	120020	10	1			
5	2015-06-22	1924-04-02	2015-06-12	12	120020	91F	6	4	12	1	120020	10	1		
6	2015-03-23	2000-08-14	2015-03-17	12	120020	14F	6NA	12	1	120020	10	2			
7	2015-03-19	1997-12-13	2015-03-11	12	120020	17F	6	4	12	1	120020	10	2		
8	2015-03-30	2012-12-23	2015-03-20	12	120020	2F	6	4	12	1	120020	10	2		
9	2015-03-23	1991-10-20	2015-03-18	12	120020	23F	6NA	12	1	120020	10	2			
10	2015-03-19	1981-04-15	2015-03-12	12	120020	33F	5	1	12	1	120020	10	2		
11	2015-03-05	1958-01-31	2015-03-01	12	120020	57F	6NA	12	1	120020	10	2			
12	2015-03-05	1995-11-28	2015-02-25	12	120020	19F	6	4	12	1	120020	10	2		
13	2015-03-05	1976-06-01	2015-02-28	12	120020	38F	6	4	12	1	120020	10	2		
14	2015-03-09	1958-06-01	2015-03-05	12	120020	56F	6NA	12	1	120020	10	2			
15	2015-03-05	1990-09-01	2015-02-26	12	120020	24F	6	4	12	1	120020	10	2		
16	2015-02-13	2003-05-19	2015-02-09	12	120020	11F	5	4	12	1	120020	10	2		
17	2015-02-20	1996-06-19	2015-02-15	12	120020	18F	5	4	12	1	120020	10	2		
18	2015-02-26	1950-02-02	2015-02-20	12	120020	65F	6NA	12	1	120020	10	2			
19	2015-02-20	2008-12-03	2015-02-13	12	120020	6F	6	4	12	1	120020	10	2		
20	2015-02-24	1950-02-02	2015-02-18	12	120020	65F	5	4	12	1	120020	10	2		
21	2015-02-26	2008-12-03	2015-02-12	12	120020	6F	6	4	12	1	120020	10	2		
22	2015-02-26	1998-07-24	2015-02-21	12	120020	16F	6	4	12	1	120020	10	2		
23	2015-02-13	1961-12-03	2015-02-10	12	120020	53F	5	4	12	1	120020	10	2		
24	2015-02-25	1966-07-09	2015-02-16	12	120020	48F	5	4	12	1	120020	10	2		
25	2015-02-11	2006-07-10	2015-02-05	12	120020	8F	6NA	12	1	120020	10	2			
26	2015-02-24	1973-09-30	2015-02-19	12	120020	41F	6	4	12	1	120020	10	2		
27	2015-02-10	1981-12-30	2015-01-09	12	120020	33F	5	4	12	1	120020	10	1		
28	2015-02-09	1996-09-14	2015-02-03	12	120020	18F	6	4	12	1	120020	10	2		
29	2015-02-11	1968-09-25	2015-02-06	12	120020	16F	6	4	12	1	120020	10	2		
30	2015-02-02	1997-01-17	2015-01-23	12	120020	18F	5	4	12	1	120020	10	2		
31	2015-02-04	1959-06-14	2015-01-30	12	120020	55F	6	4	12	1	120020	10	2		
32	2015-02-04	1947-06-07	2015-02-02	12	120020	67F	6	4	12	1	120020	10	2		
33	2015-02-04	1977-06-03	2015-01-30	12	120020	37F	6NA	12	1	120020	10	2			



Clean health data

Comparison of Dengue Cases in BOYACA, CASANARE



Load

Load from common file formats (csv, rds, xlsx, json, dbf, dbc)

Clean

- Standardize names and values
- Convert dates
- Remove missing or invalid records

Filter

- Select:
 - Date ranges
 - Demographics
 - Numeric thresholds

Aggregate

- Summarize cases over:
 - space
 - time
 - other relevant variables

Code-experienced user

```
d4h_load()
```

```
d4h_clean()
```

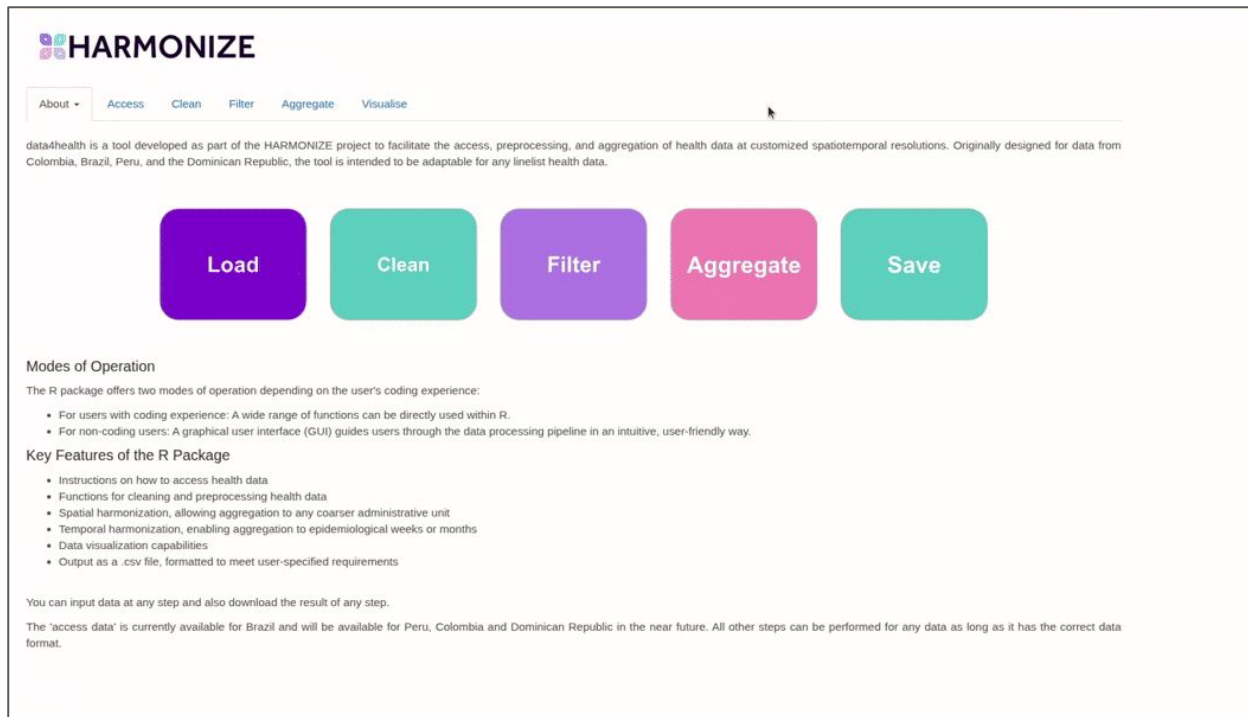
```
d4h_filter()
```

```
d4h_aggregate()
```

```
d4h_save()
```

Non-code-experienced user

```
d4h_ui()
```

A screenshot of the HARMONIZE web application interface. At the top, there's a navigation bar with a logo and the word 'HARMONIZE'. Below it, a menu bar contains links: 'About', 'Access', 'Clean', 'Filter', 'Aggregate', and 'Visualise'. A paragraph of text describes the tool's purpose: 'data4health is a tool developed as part of the HARMONIZE project to facilitate the access, preprocessing, and aggregation of health data at customized spatiotemporal resolutions. Originally designed for data from Colombia, Brazil, Peru, and the Dominican Republic, the tool is intended to be adaptable for any list of health data.' Below this text is a horizontal pipeline of five colored buttons: 'Load' (purple), 'Clean' (teal), 'Filter' (purple), 'Aggregate' (pink), and 'Save' (teal). Further down, a section titled 'Modes of Operation' explains that the R package offers two modes: one for coding experience and another for non-coding users using a GUI. This is followed by 'Key Features of the R Package' which lists capabilities like accessing health data, cleaning/preprocessing, spatial/temporal harmonization, visualization, and output formatting. The bottom section states that users can input data at any step and download results, and mentions that 'access data' is currently available for Brazil and will be available for Peru, Colombia, and the Dominican Republic in the near future.



An R package designed to obtain, transform and export climate data for their use in epidemiological analyses and other types of applications.



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CLIMATE RESEARCHER



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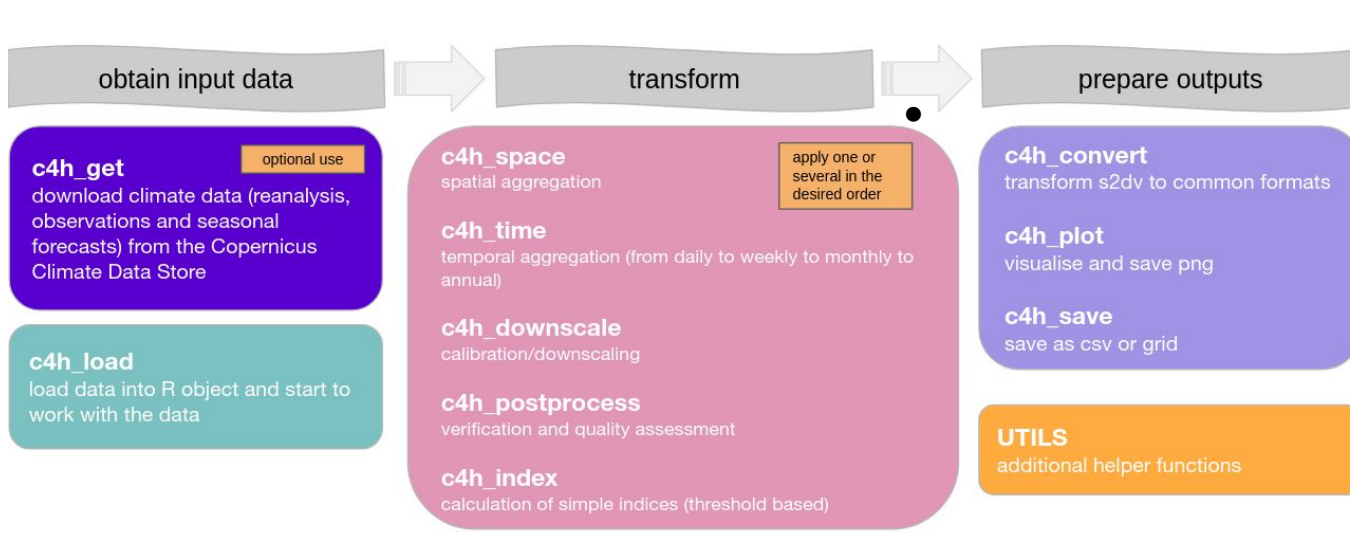


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SOFTWARE ENGINEER



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Expected release date:
July 2026



Seasonal forecasts available from Copernicus:

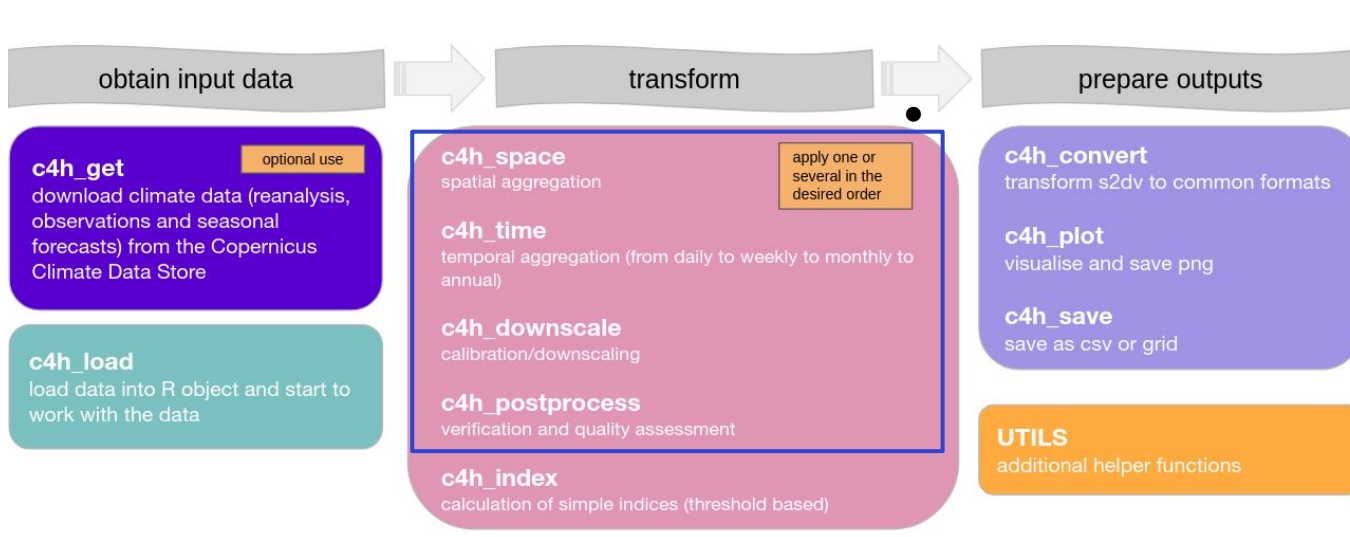
- Spatial resolution: $1^{\circ} \times 1^{\circ}$ (~100km)
- System: ECMWF-s5

Reanalysis available from Copernicus:

- ERA5 1979-present $0.25^{\circ} \times 0.25^{\circ}$ (~28km)
- ERA5-Land 1950-present $0.1^{\circ} \times 0.1^{\circ}$ (~10km)

Ground based stations:

- Point data
- Daily or sub-daily temporal resolution



Seasonal forecasts available from Copernicus:

- Spatial resolution: $1^{\circ} \times 1^{\circ}$ (~100km)
- System: ECMWF-s5

Reanalysis available from Copernicus:

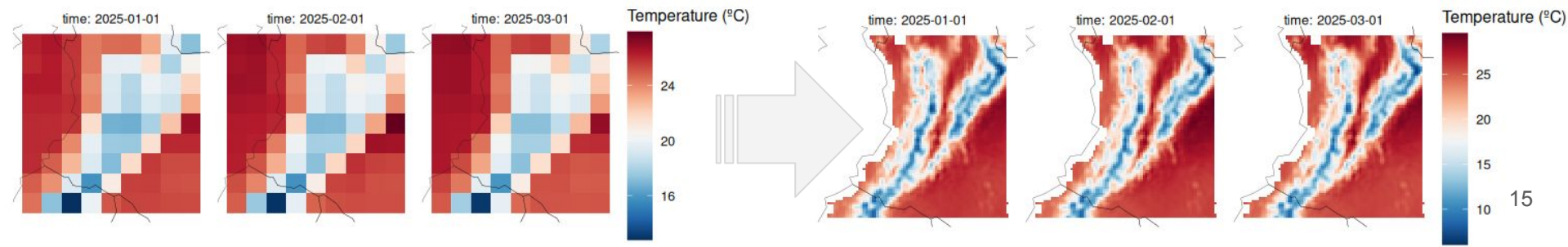
- ERA5 1979-present $0.25^{\circ} \times 0.25^{\circ}$ (~28km)
- ERA5-Land 1950-present $0.1^{\circ} \times 0.1^{\circ}$ (~10km)

Ground based stations:

- Point data
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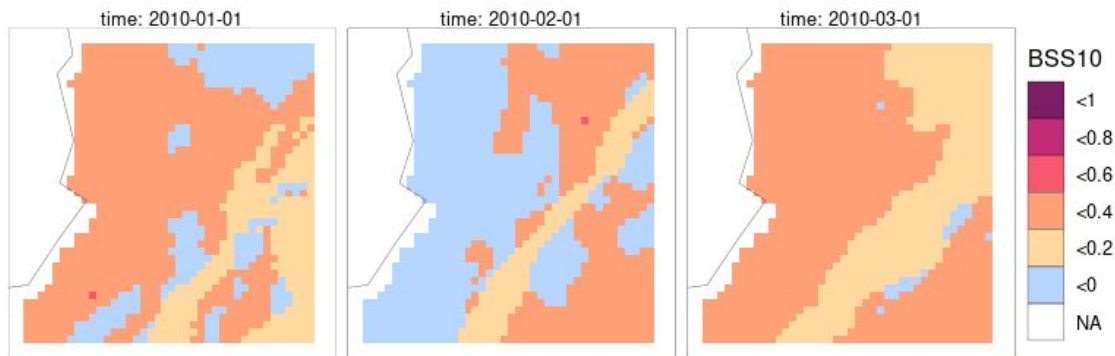
We use hindcasts and reanalysis ($\sim 10\text{km}$) to downscale and calibrate seasonal forecasts ($\sim 100\text{km}$). Available methods include:

- Interpolation (bilinear, bicubic, nearest neighbour, conservative)
- Interpolation + bias correction (Gaussian, EVMOS, quantile mapping)
- Interpolation + linear regression (cell-wise, focal)
- Analogs (K-NNs)



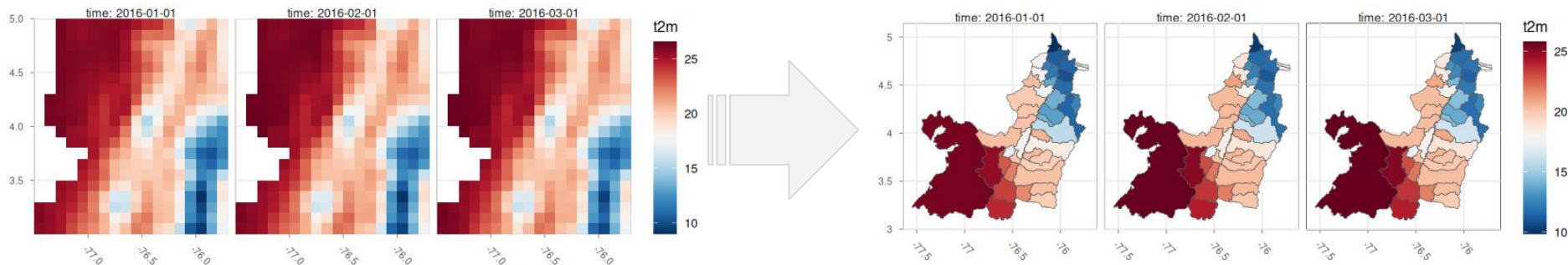
We assess the quality of the seasonal forecasts using hindcasts and reanalysis data and compute skill metrics to compare it to the climatologies:

- Brier Skill Score (BSS)
- Ranked probability Skill Score (RSS)
- Continuous Ranked Probability Skill Score (CRPSS)
- + Many others (MSS, MAE, ROC, RMS, AbsBias)



We aligned the spatiotemporal resolution of the climate data to the spatio-temporal resolution of the epidemiological data:

- Ensemble aggregations
- Grid-to-polygon aggregations
- Hourly/daily to weekly/monthly aggregations





An R package for exploratory analysis of temporal and spatio-temporal health data including case counts, incidence rates, and covariates.



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VISITING RESEARCHER



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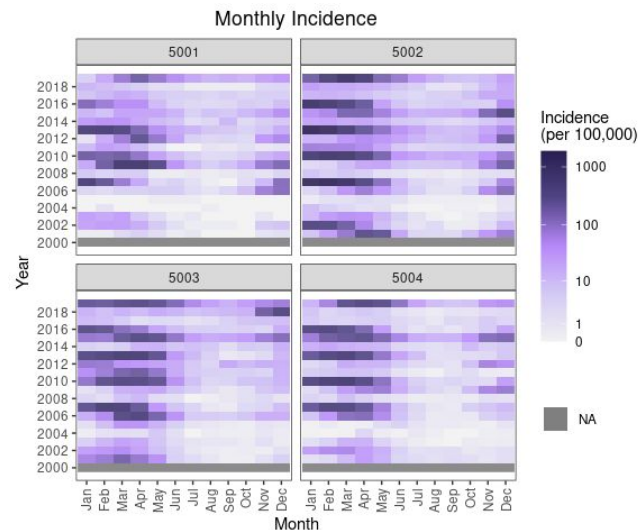
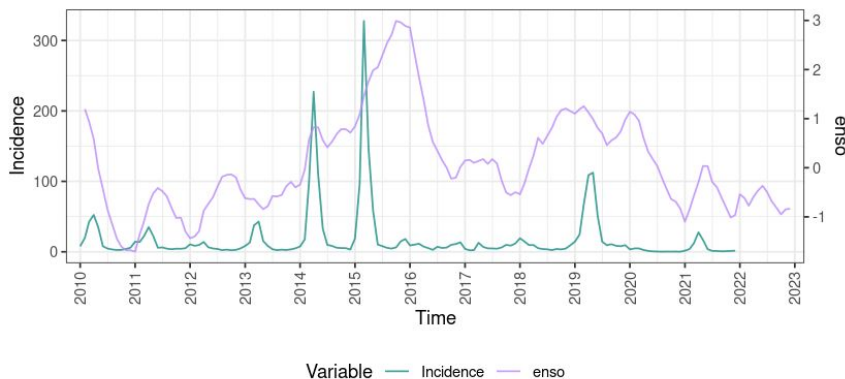


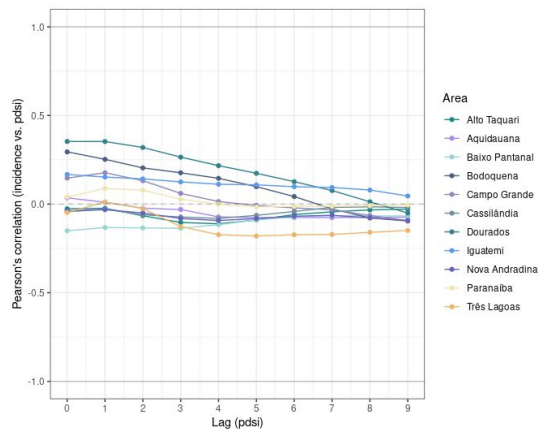
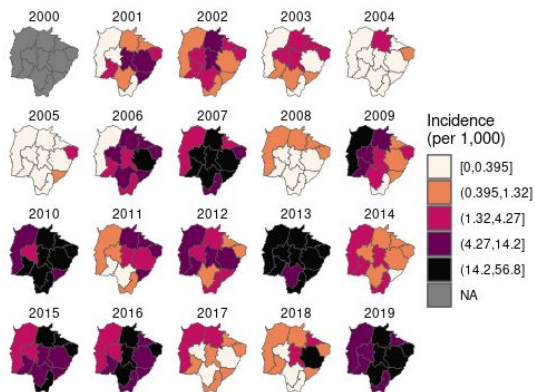
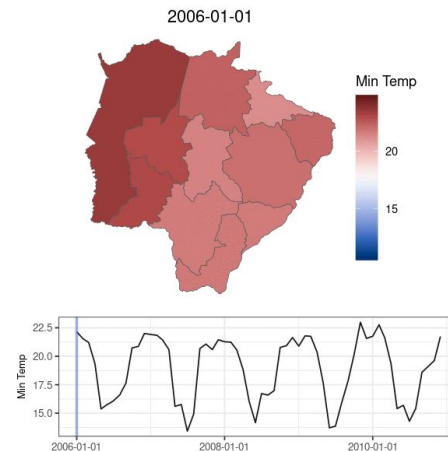
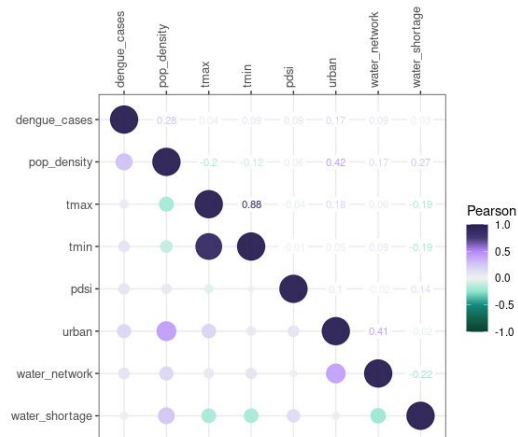
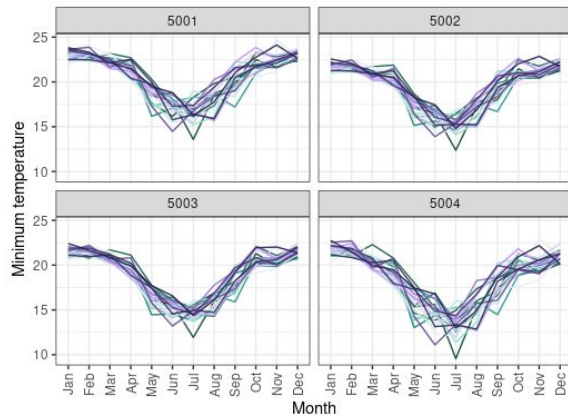
PROF RACHEL LOWE
LEADING RESEARCHER

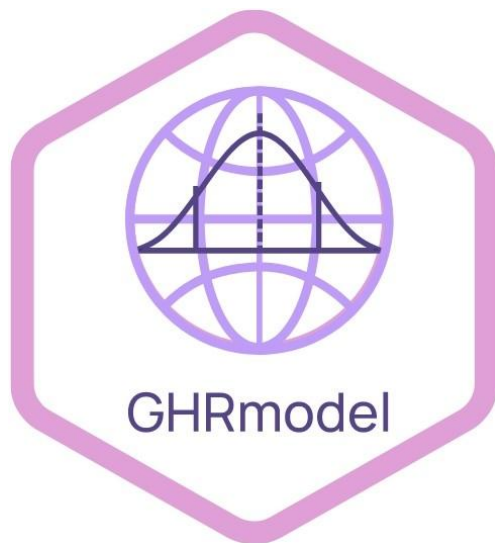
Version 0.2.1 on CRAN

Easy visualization of spatio-temporal disease count, incidence and covariate tabular data:

- Explore spatio-temporal structure (seasonality, interannual, spatial patterns)
- Explore relationships with covariates (lags, associations)







An R package for modeling health outcomes using Bayesian hierarchical spatio-temporal models with complex covariate effects in the R-INLA framework.



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Version 0.1.1 on CRAN

$$Y_{i,t} | \mu_{i,t}, \kappa \sim \text{NegBin}(\mu_{i,t}, \kappa)$$

$$\log(\mu_{i,t}) = \alpha + \log(P_{i,a[t]}) + \sum_k \beta_k X_{i,t,k} + \gamma_{i,a[t]} + \delta_{i,m[t]}$$

Explanatory variables:

- Linear
- Non-linear
- Random slopes
- Cross-basis matrices (DLNMs)
- Interactions
- ...

Random effects:

- Seasonal
- Interannual
- Spatial
- ...

INLA

- Statistical models?
Transparency
- Bayesian?
Uncertainty quantification
- INLA?
Fast Bayesian inference

Transform
covariates

Define
formulas

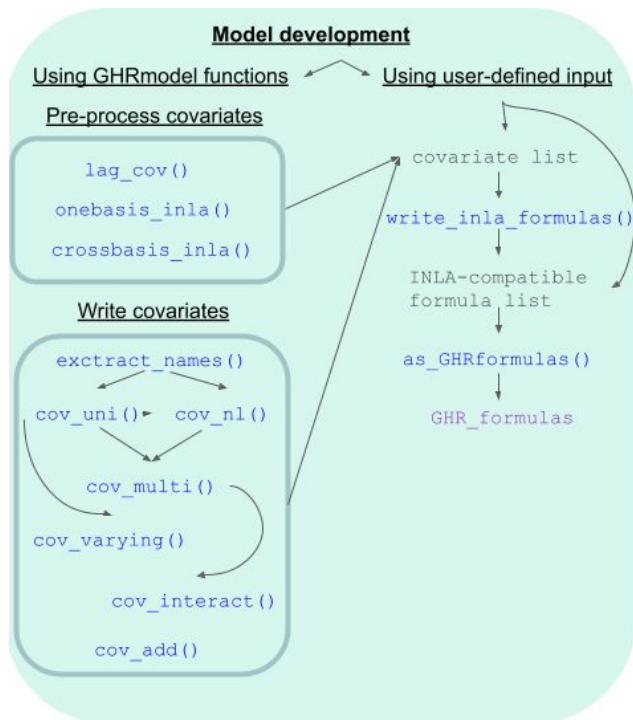
Fit models

Model
evaluation

Transform
covariates



Define
formulas



- Helpers to prepare long lists of formulas to be fitted in INLA including linear and non-linear effects, interactions, and random slopes.
- Helpers to specify complex random effects.
- Prepare one- and cross-basis functions.



We fit INLA models in batches which we can then easily compare to each other:

```
model_list <- fit_models(  
  formulas = formulas_cov_list_ghr, # GHRformulas object  
  data = data,                      # Data to fit the models  
  family = "nbinomial",             # Negative binomial likelihood  
  name = "mod",                     # Label prefix for each model  
  offset = "population",            # Offset variable to account for population size  
  control_compute = list(  
    config = FALSE,                 # Extract from INLA summary.fitted.values  
    vcov = FALSE                    # Do not return variance-covariance matrix  
  ),  
  nthreads = 8                      # Use 8 threads for parallel computation  
)
```

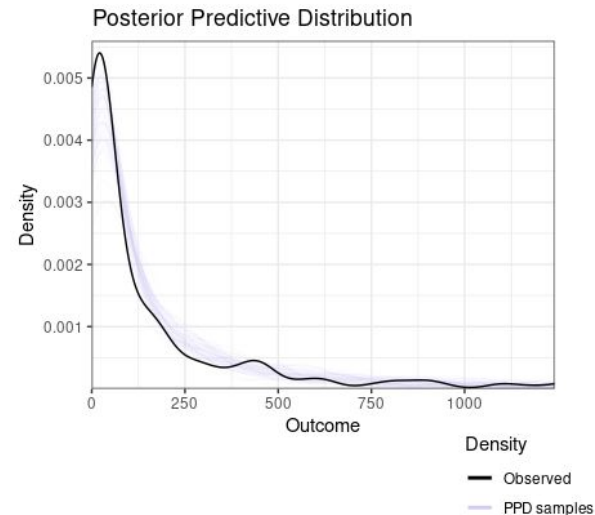
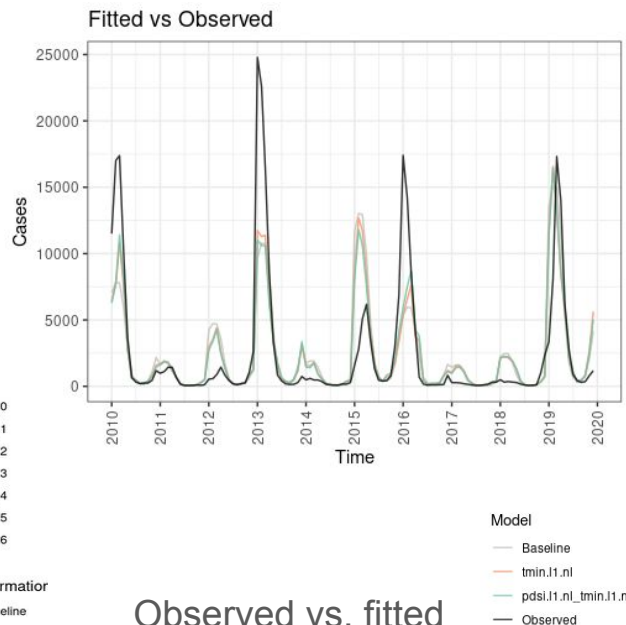
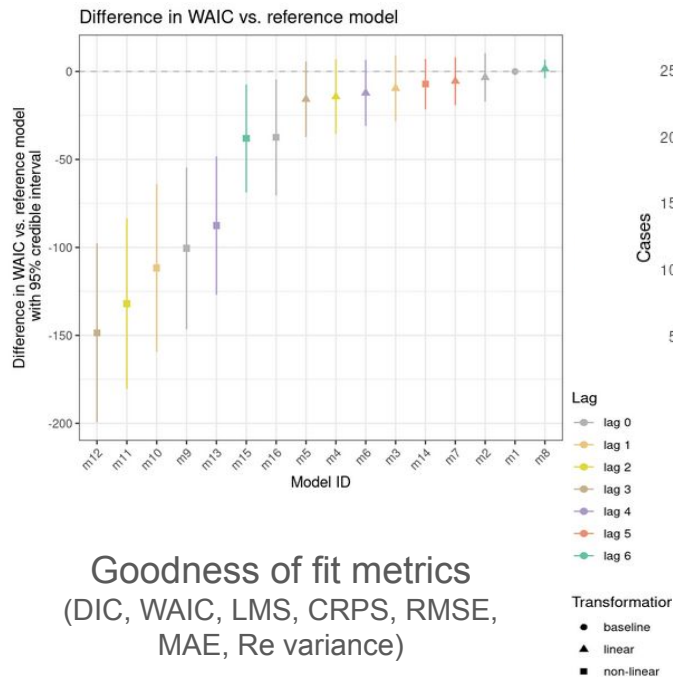
And extract a series of (option) outputs to be used when examining the models

Transform
covariates

Define
formulas

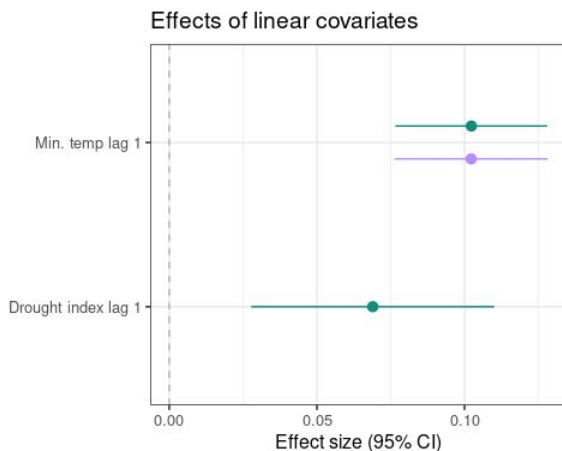
Fit models

Model
evaluation



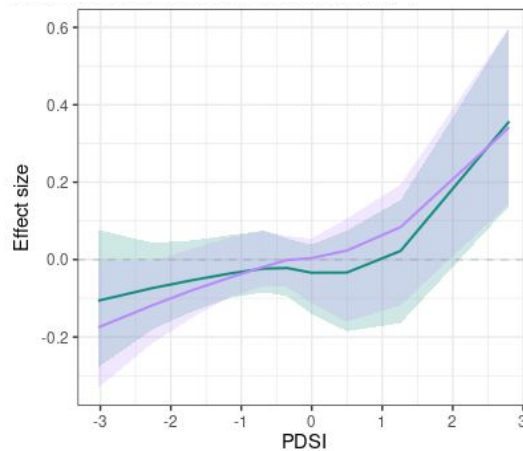


Evaluate effects: linear, non-linear, and DLNMs posterior estimates



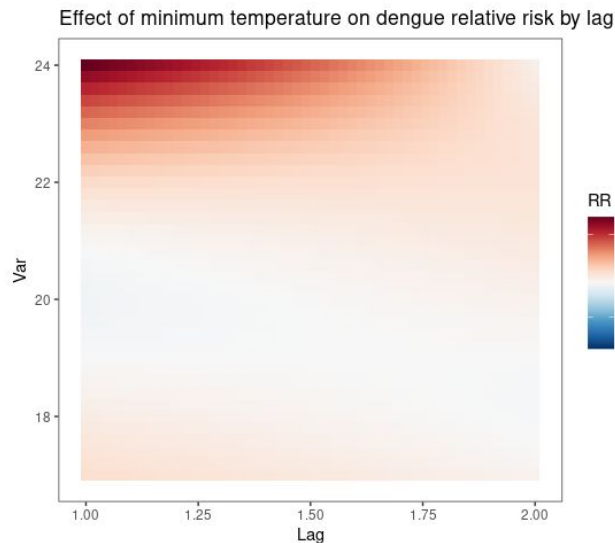
Model

- mod2
- mod4



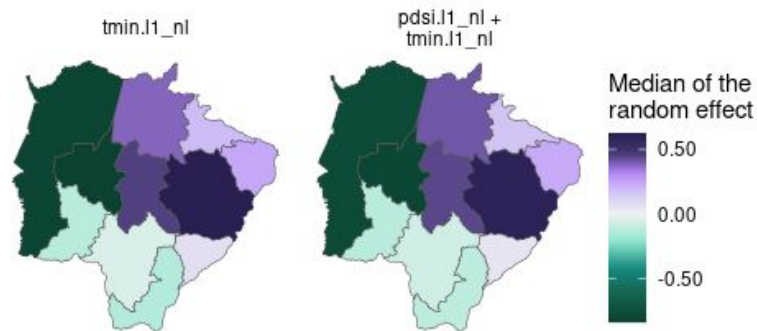
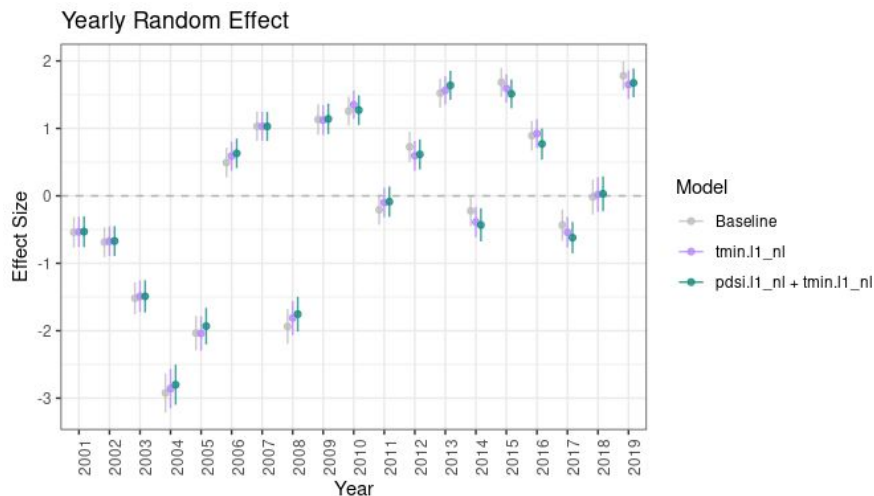
Model

- pdsi.l1_ni
- pdsi.l1_ni + tmin.l1_ni





Evaluate effects: random effects posterior estimates





An R package designed for computing out-of-sample predictions, enabling cross-validation analysis and the evaluation of predictive performance for INLA models.



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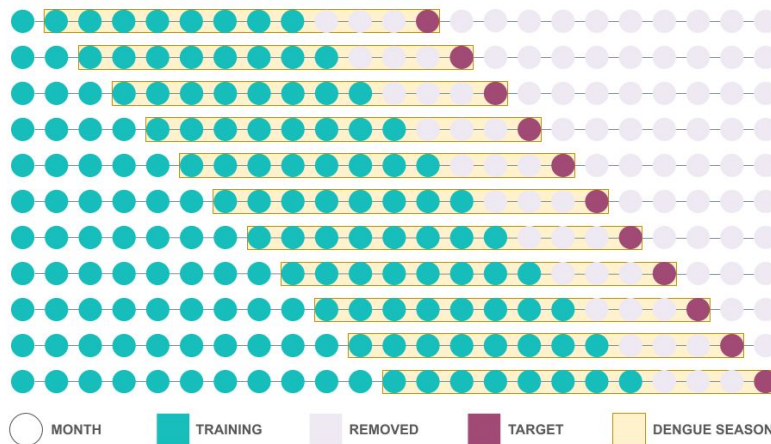
Expected release date:
December 2026

Cross-validation



Predictive
performance

Expanding window cross-validation



Source: Fletcher et al. (2025)

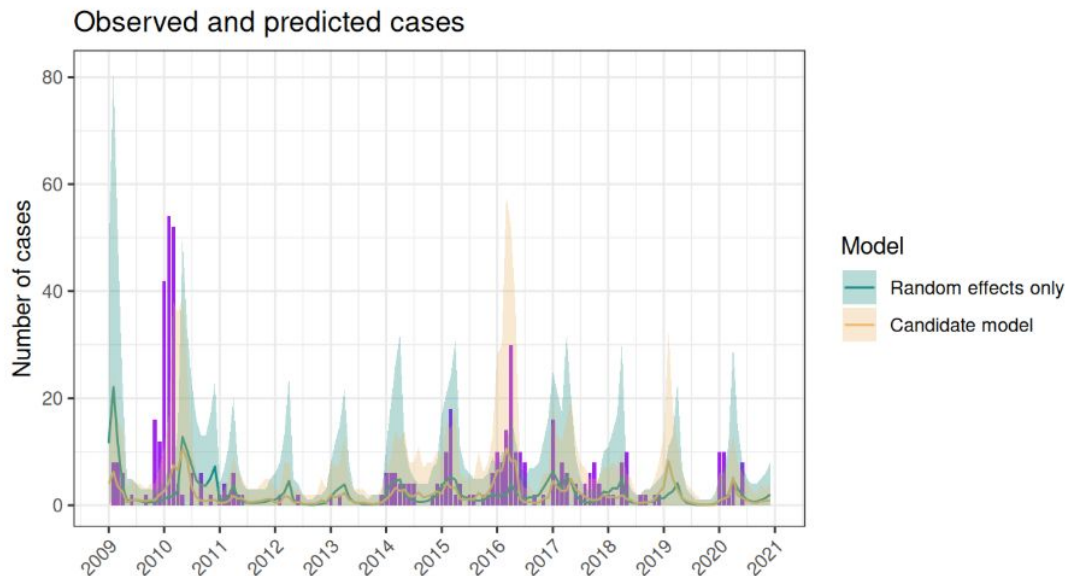
Cross-validation



Predictive
performance

Computes forecast accuracy metrics
for case counts prediction:

- Continuous Ranked Probability Score (CRPS)
- Mean Absolute Error (MAE)
- Root Mean Square Error (RMSE).



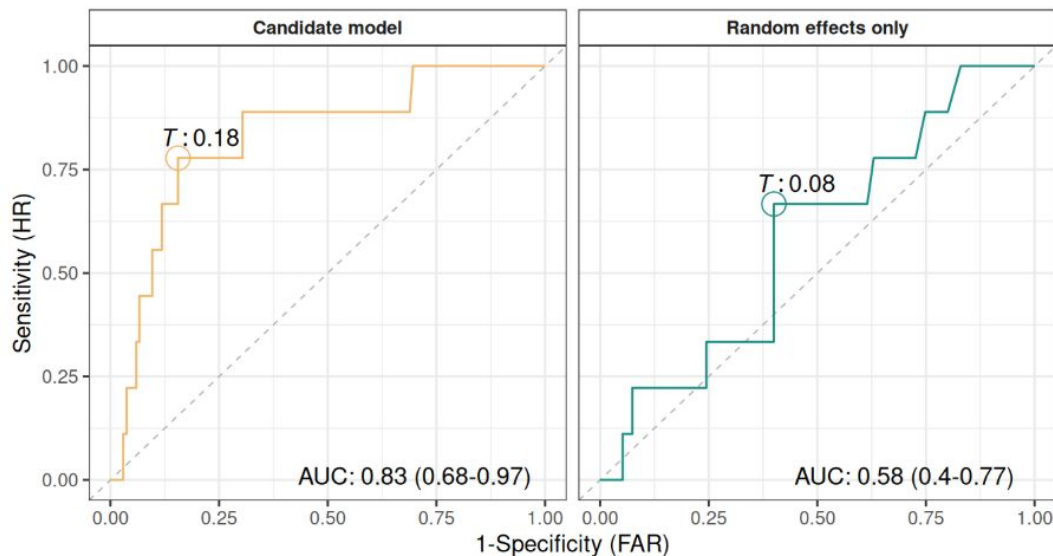
Cross-validation

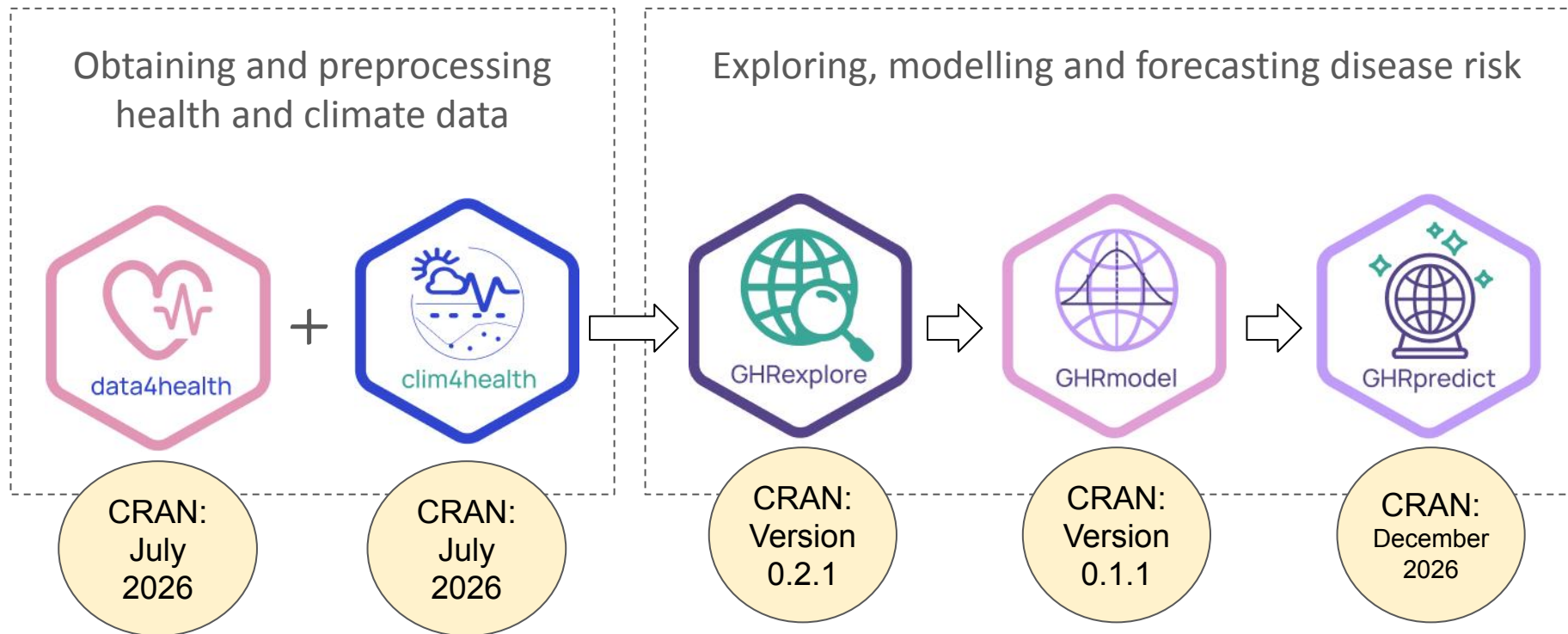


Predictive
performance

Computes forecast accuracy metrics for outbreak detection:

- Multiple thresholds to define outbreaks.
- Metrics: ROC, Brier, Confusion matrix
- Optimal trigger outbreak probability based on ROC curves.





GHRtools Home data4health clim4health GHRexplore GHRmodel GHRpredict About GHR Q

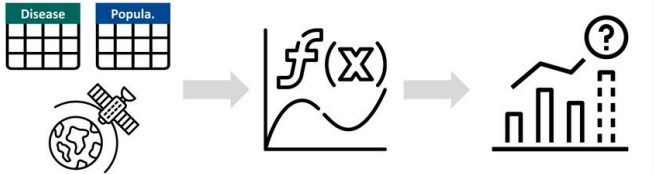
GHRtools

A suite of R packages for climate and health workflows created by the Global Health Resilience (GHR) group at the Barcelona Supercomputing Center.



From health and climate data to disease forecasts

GHRtools is a suite of R packages that support **climate and health workflows** by streamlining health and climate data collection and harmonization, exploratory analysis, Bayesian modelling and forecasting.



Harmonised health and climate data Bayesian model Disease prediction

<https://bsc-es.github.io/GHRtools>



- Documentation
- Vignettes and tutorials
- Software manuals

Thank you!

