

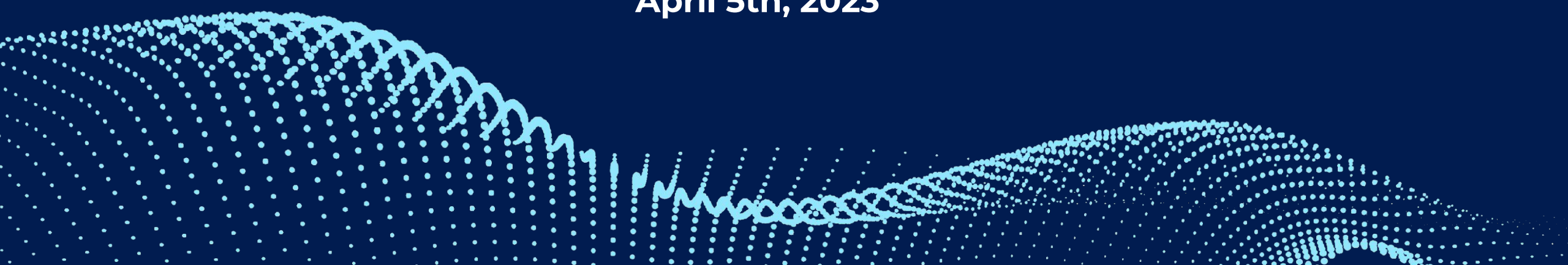
Epiverse
TRACELAC



Pontificia Universidad
JAVERIANA
Bogotá

Serofoi 0.0.9

April 5th, 2023



serofoi 0.0.9

Authors:

- Zulma M. Cucunubá
- Nicolás Tórres
- Benjamin Lambert
- Pierre Nouvellet

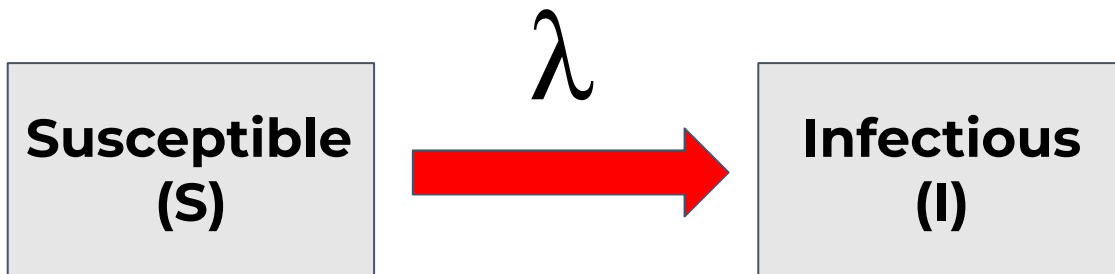


Contributors:

- Geraldine Gómez
- Jaime A. Pavlich-Mariscal
- Miguel Gámez

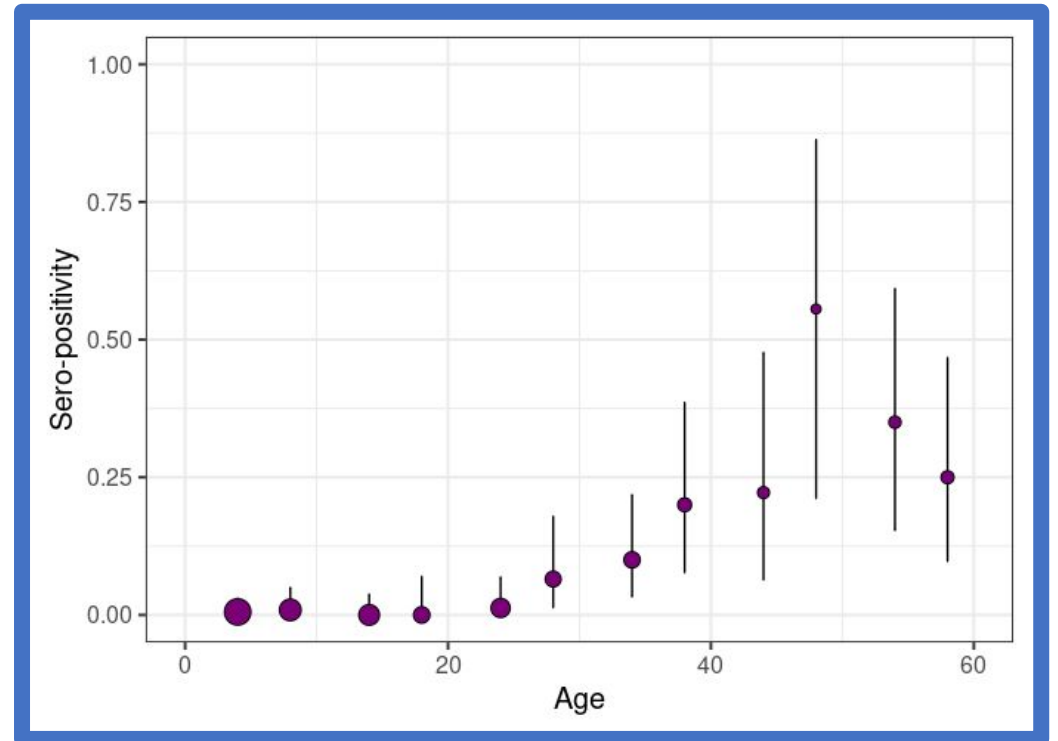
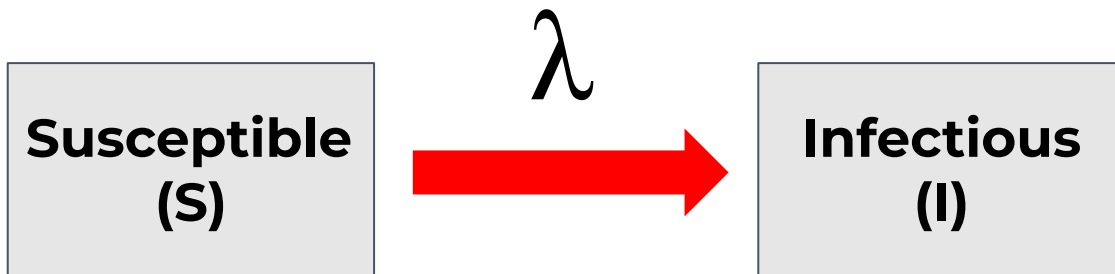
serofoi 0.0.9

An R package for estimating the **Force-of-Infection (Fol)**



serofoi 0.0.9

An R package for estimating the **Force-of-Infection (Fol)** from **age-disaggregated** population-based cross-sectional serosurveys.



serofoi 0.0.9



Usage:

Infectious diseases for which IgG antibodies can be measured.

Serosurvey Criteria for serofoi



Serosurvey Criteria for serofoi



- Are population-based cross-sectional surveys (not hospital-based).

Serosurvey Criteria for serofoi



- Are population-based cross-sectional surveys (not hospital-based).
- Specify individuals' age or age group.

Serosurvey Criteria for serofoi



- Are population-based cross-sectional surveys (not hospital-based).
- Specify individuals' age or age group.
- Indicate diagnostic test(s) used. The current version of **serofoi** only applies to IgG antibodies.

Serosurvey Criteria for serofoi



- Are population-based cross-sectional surveys (not hospital-based)
- Specify individuals' age or age group
- Indicate diagnostic test(s) used. The current version of **serofoi** only applies to IgG antibodies
- Identify the date (year) and place of sample collection

Biological Assumptions of serofoi



Biological Assumptions of serofoi



- No sero-reversion.

Biological Assumptions of serofoi



- No sero-reversion.
- No age-dependency.

Biological Assumptions of serofoi



- No sero-reversion.
- No age-dependency.
- No impact from migration processes.

Biological Assumptions of serofoi

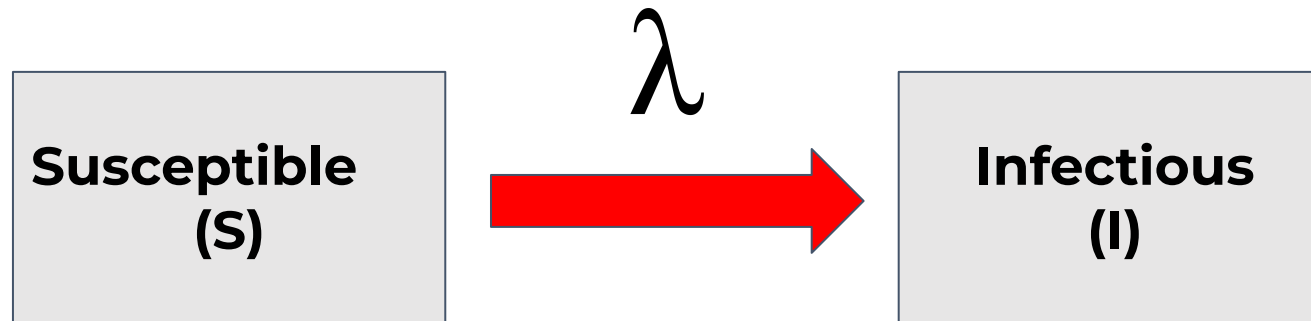


- No sero-reversion.
- No age-dependency.
- No impact from migration processes.
- No differences in the mortality rate of infected versus susceptible individuals.

Force of Infection (FoI)



Rate at which susceptible individuals exposed to a pathogen become infected.





Constant vs Time-varying Fol:

- The Fol is often incorrectly assumed to be constant over time.



Constant vs Time-varying Fol:

- The Fol is often incorrectly assumed to be constant over time.

serofoi allows both

- Constant Fol
- Time-varying Fol



Constant vs Time-varying Fol:

Model Option	Description and usage
constant	Constant Fol
tv_normal	Time-varying normal Fol: slow change in Fol
tv_normal_log	Time-varying normal-log Fol: fast epidemic change in Fol

https://epiverse-trace.github.io/serofoi/articles/foi_models.html

Getting started with *serofoi*



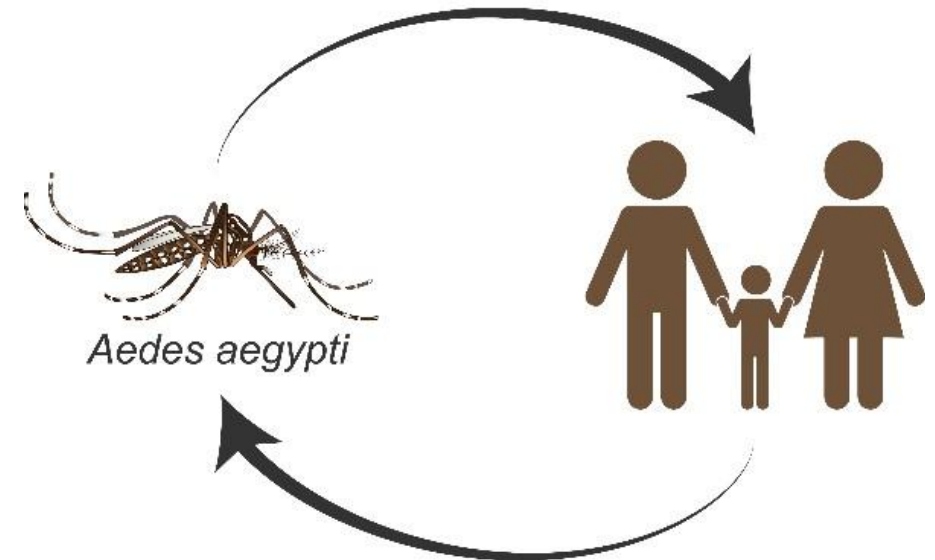
Case study 1: Chikungunya (fast-spreading epidemic)



Case study 1: Chikungunya (fast-spreading epidemic)



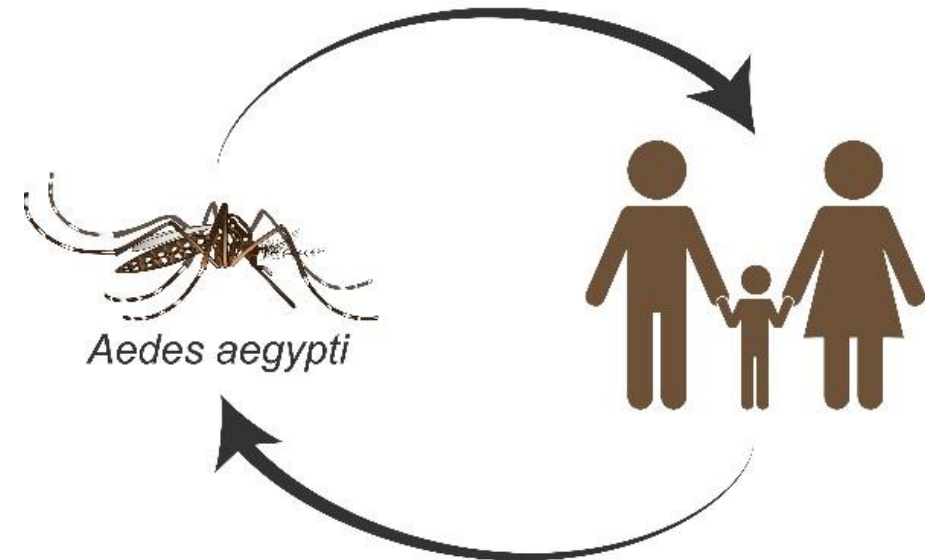
- Transmitted mainly by the *Aedes aegypti* mosquito



Case study 1: Chikungunya (fast-spreading epidemic)



- Transmitted mainly by the *Aedes aegypti* mosquito
- Originally endemic to Africa and Asia
- Arrives to the Americas on 2013 with no prior immunity in the population



Case study 1: Chikungunya (fast-spreading epidemic)



Methodological challenge:

How is it best to untangle the endemic and epidemic patterns of Chikungunya?

Case study 1: Chikungunya (fast-spreading epidemic)



```
# Load and prepare data  
data("chik2015")  
chik2015p <- prepare_serodata(chik2015)
```

population-based study conducted in Bahia, Brazil in
October-December 2015.

Case study 1: Chikungunya (fast-spreading epidemic)



```
# Implementation of the models
m1_chik <- run_seromodel(serodata = chik2015p,
                        foi_model = "constant",
                        n_iters = 1000,
                        n_thin = 2)

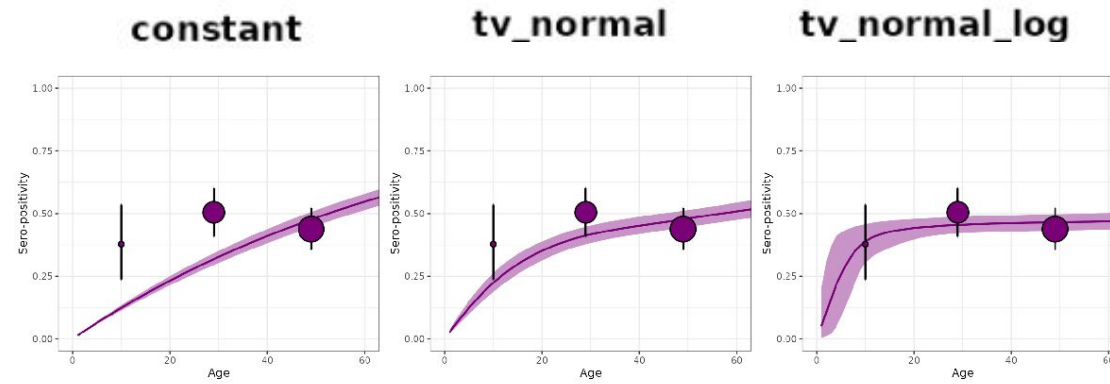
m2_chik <- run_seromodel(serodata = chik2015p,
                        foi_model = "tv_normal",
                        n_iters = 1500,
                        n_thin = 2)

m3_chik <- run_seromodel(serodata = chik2015p,
                        foi_model = "tv_normal_log",
                        n_iters = 1500,
                        n_thin = 2)
```

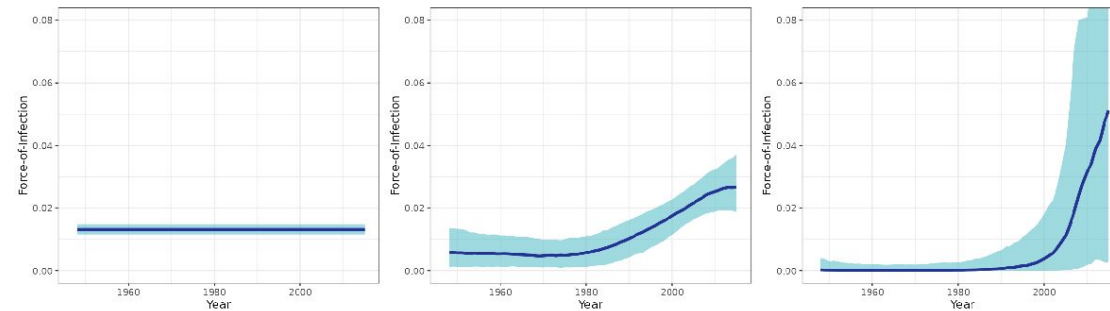
Case study 1: Chikungunya (fast-spreading epidemic)



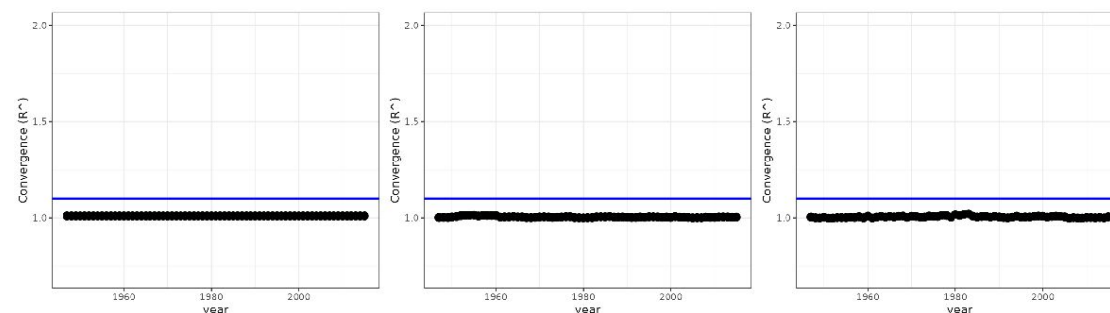
Seroprevalence
Fitting



Fol estimate



Convergence
criteria



Case study 1: Chikungunya (fast-spreading epidemic)



foi_model: constant
dataset: BRA 2015(S019)
elpd: -39.57
se: 8.48
converged: Yes

foi_model: tv_normal
dataset: BRA 2015(S019)
elpd: -20.76
se: 1.98
converged: Yes

foi_model: tv_normal_log
dataset: BRA 2015(S019)
elpd: -12.58
se: 0.81
converged: Yes

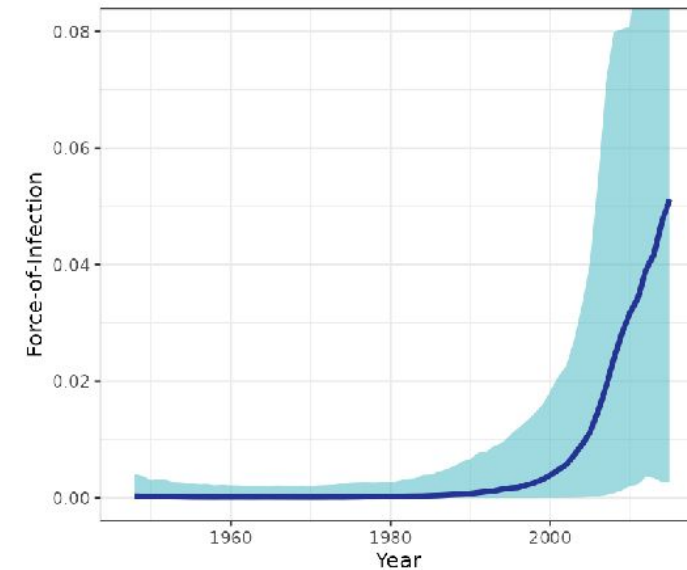
Case study 1: Chikungunya (fast-spreading epidemic)



foi_model: constant
dataset: BRA 2015(S019)
elpd: -39.57
se: 8.48
converged: Yes

foi_model: tv_normal
dataset: BRA 2015(S019)
elpd: -20.76
se: 1.98
converged: Yes

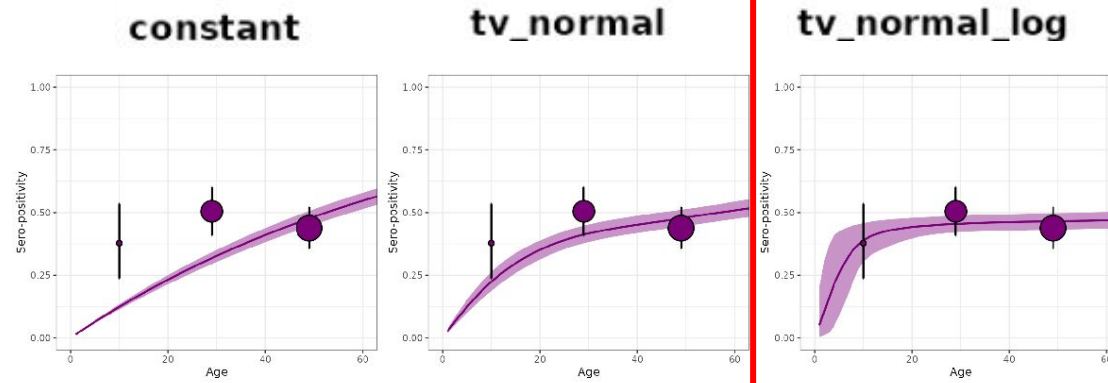
foi_model: tv_normal_log
dataset: BRA 2015(S019)
elpd: -12.58
se: 0.81
converged: Yes



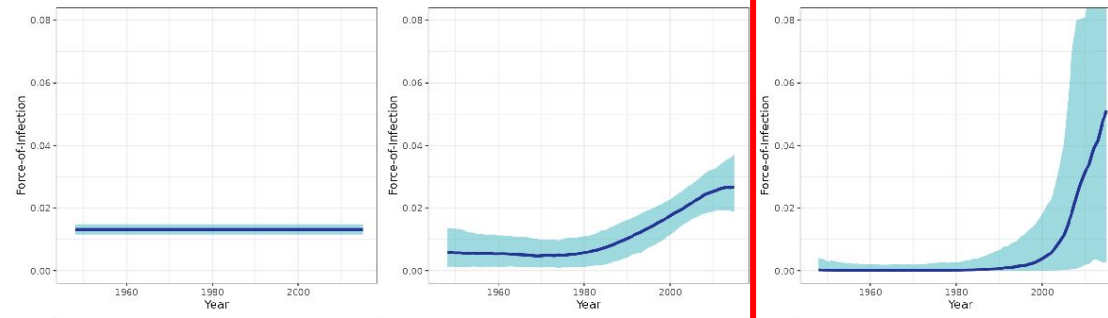
Case study 1: Chikungunya (fast-spreading epidemic)



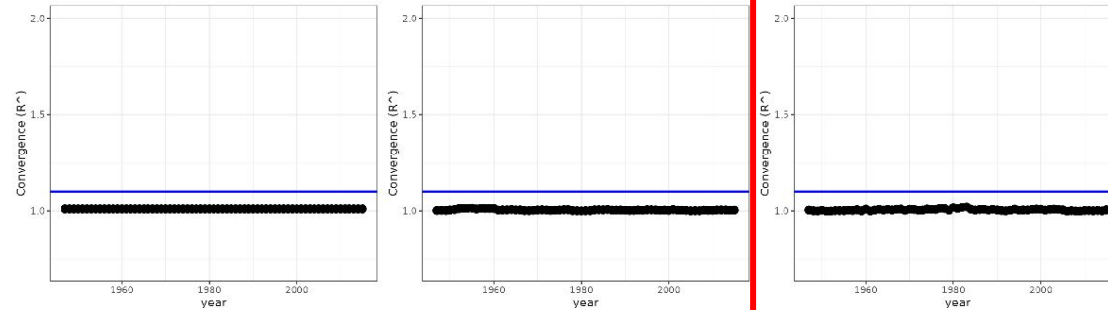
Seroprevalence
Fitting



Fol estimate



Convergence
criteria



Case study 2: Chagas Disease



Case study 2: Chagas Disease



- *Trypanosoma cruzi*
- Endemic to Latin America
- Transmitted by triatomine bugs



Case study 2: Chagas Disease



Control: Insecticide spraying has been extensively implemented in Latin America since the 1970s-80s.



Case study 2: Chagas Disease



Methodological challenge:

Estimate the historical effectiveness of the control strategies across endemic areas?

Case study 2: Chagas Disease



```
# Load and prepare data
data("chagas2012")
chagas2012p <- prepare_serodata(chagas2012)

# Implementation of the models
m1_cha <- run_seromodel(serodata = chagas2012p,
                       foi_model = "constant",
                       n_iters = 800)
m2_cha <- run_seromodel(serodata = chagas2012p,
                       foi_model = "tv_normal",
                       n_iters = 800)
```

Case study 2: Chagas Disease

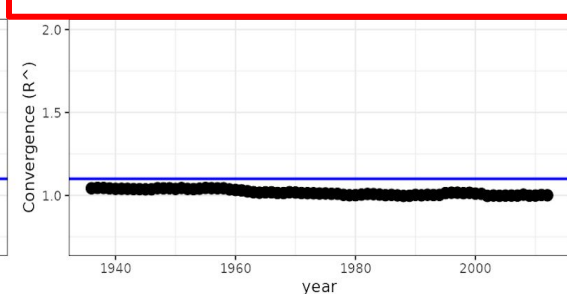
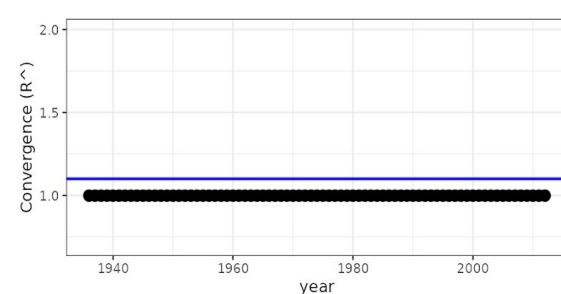
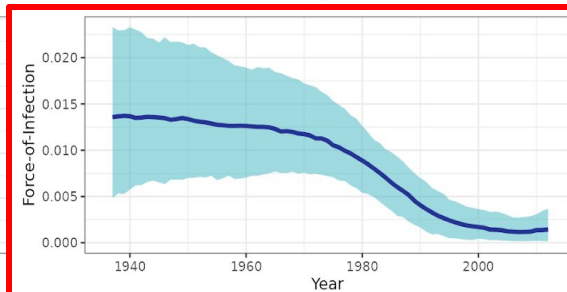
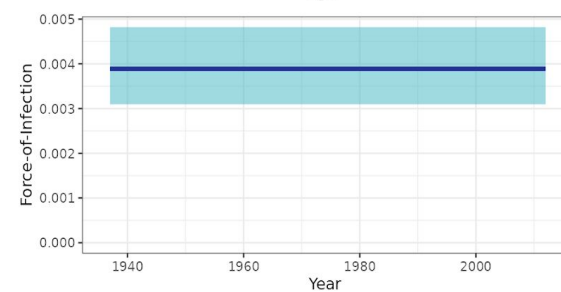
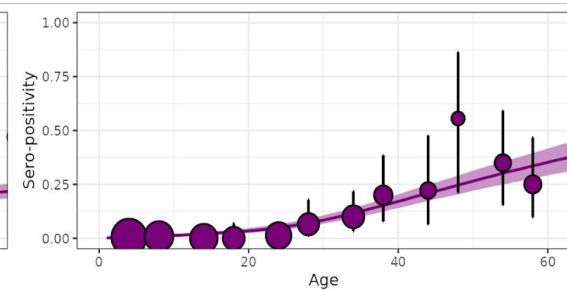
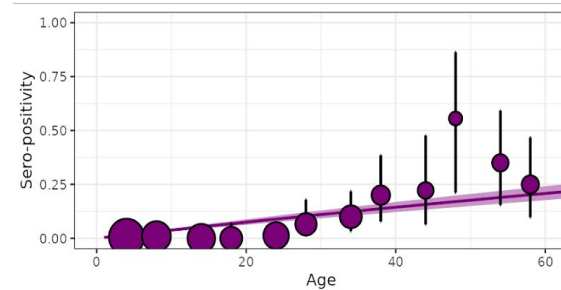


foi_model: constant
elpd: -92.64
se: 6.41
converged: Yes

foi_model: tv_normal
elpd: -74.52
se: 5.92
converged: Yes

```
# Load and prepare data
data("chagas2012")
chagas2012p <- prepare_serodata(chagas2012)
```

```
# Implementation of the models
m1_cha <- run_seromodel(serodata = chagas2012p,
  foi_model = "constant",
  n_iters = 800)
m2_cha <- run_seromodel(serodata = chagas2012p,
  foi_model = "tv_normal",
  n_iters = 800)
```



Conclusions



- **serofoi** Fol estimations in different real case scenarios
- **serofoi** is able to recover the endemic or epidemic trends of different diseases
- Bayesian comparison criteria can be used to identify trends on the Fol
- Future versions of **serofoi** might include additional models, visualization tools and further model comparison criteria



Contribute in serofoi:

Contributions are welcome via pull requests, taking into account the code of conduct.

GitHub: <https://github.com/epiverse-trace/serofoi>

Website: <https://epiverse-trace.github.io/serofoi/>

Get in touch:

Email:

ex-ntorres@javeriana.edu.co

ORCID:

<https://orcid.org/0009-0002-8484-1298>

Thanks!

