

Phylodynamics of acute HCV infection in men having sex with men

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Disease Modeling Symposium 2019



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Two ongoing epidemics in Lyon

"Classical" hosts

- Injection drug users
- Detection during chronic infection
- HIV +

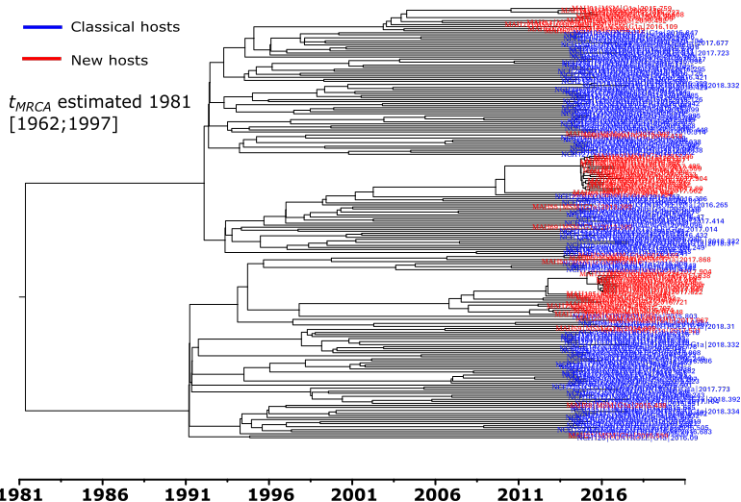
New hosts are being detected since 2000's

- Men who have sex with men (MSM)
- Detection during acute infection
- HIV - and HIV +



213 sequences (NS5B gene, 322bp)
Known sampling dates from 2011 to 2018

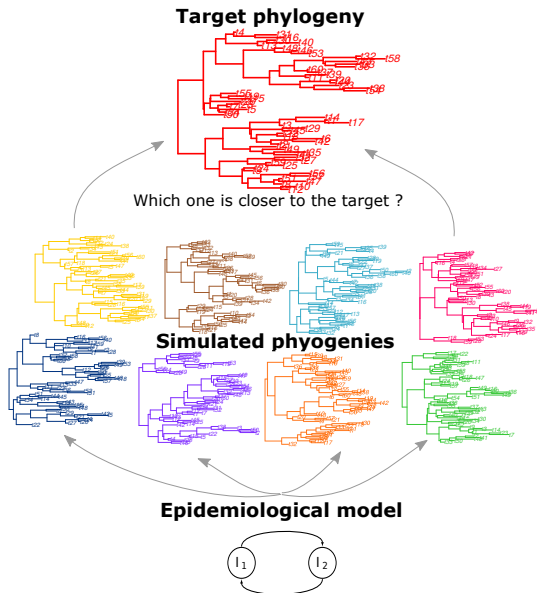
Known sampling dates from 2011 to 2018



Our aim

- Perform phylodynamics inference using two types of hosts.
- Infer the basic reproduction number (R_0) for each epidemic simultaneously.

ABC Phylogenetics



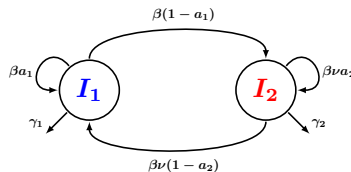
Epidemiological model

Host compartments

- I_1 : classical hosts
- I_2 : new hosts

Parameters

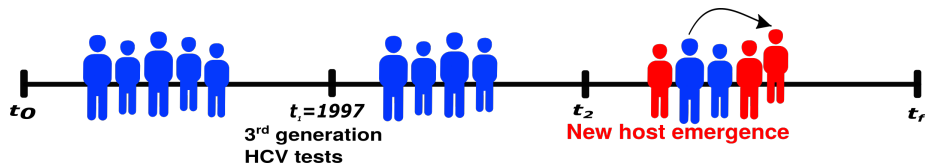
- β : transmission rate
- ν : transmission differential
- a_i : assortativity
- $1/\gamma_i$: infectious period
- $R_0 = \frac{\beta}{\gamma}$: reproduction number



$$\frac{dI_1}{dt} = a_1\beta I_1 + (1 - a_2)\nu\beta I_2 - \gamma_1 I_1$$

$$\frac{dI_2}{dt} = a_2\beta\nu I_2 + (1 - a_1)\beta I_1 - \gamma_2 I_2$$

Priors



	t_0-t_1	t_1-t_2	t_2-t_f
γ_1	Unif(0.1, 4)	-	-
γ_2	0	0	Unif(0.1, 3)
ν	0	0	Unif(1, 10)
a_i	Unif(0, 1)	-	-
R_0^1	Unif(0.9, 15)	Unif(0.1, 3)	-

$$\beta_{t1} = R_{0,t1}^1 \gamma_1$$

$$\beta_{t2} = R_{0,t2}^1 \gamma_1$$

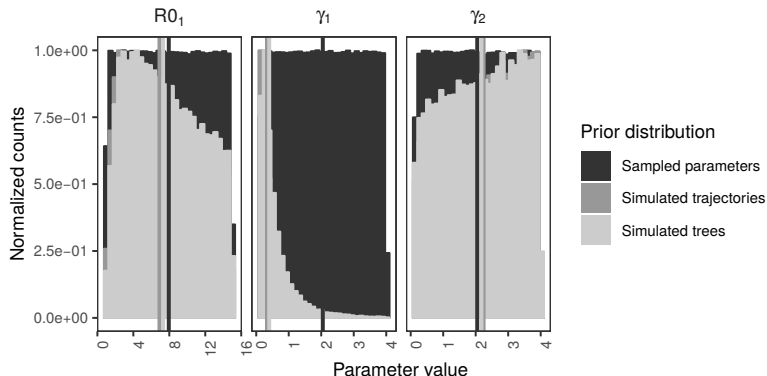
$$R_{0,t3}^2 = \frac{\beta_{t2} \nu}{\gamma_2}$$

Simulations of phylogenies

- Simulation of dated phylogenies using a two-step procedure:
 1. Simulate epidemiological trajectories given our model using Gillespie Stochastic Simulations Algorithm (SSA)
 2. Simulate phylogenies using simulated trajectories and sampling dates, going backward-in-time
- Simulator EpidSim implemented in Rcpp
- Number of phylogenies simulated (training dataset) is key

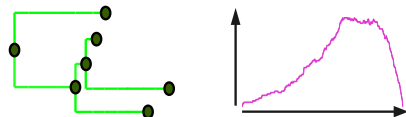
The feasibility of the simulations reduces the prior distribution

2.10^6 sets of parameters \rightarrow 61.000 simulated phylogenies



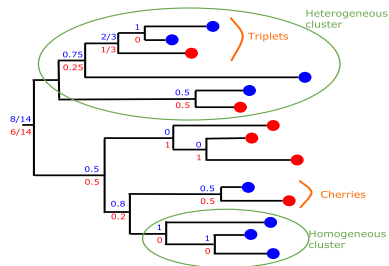
Summary statistics

- Branch lengths
- Tree topology
- LTT (lineage through-time) plot
- X-axis coordinates of the LTT plot
- Y-axis coordinates of the LTT plot



New summary statistics

- Cherries, triplets
- Branch lengths on homogeneous and heterogeneous clusters
- LTT 'label' plot

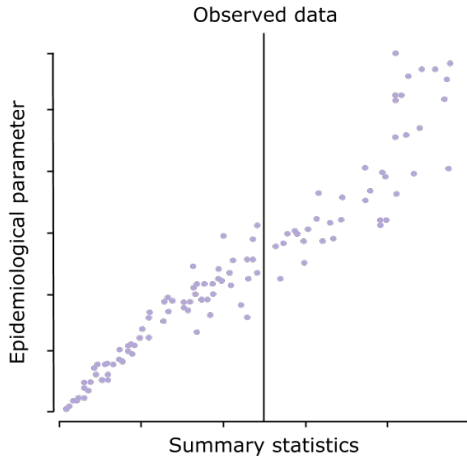


Saulnier et al. (2017, *PLOS Comput. Biol.*)

Steps for the ABC Elastic Net regression

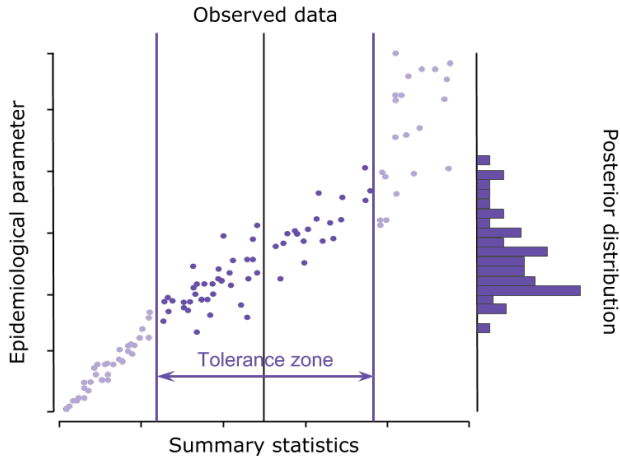
1. Test for multicollinearity using Variance Inflation Factors (VIF)
 - Stepwise variable selection
 - $VIF_j = \frac{1}{1-R_j^2}$
 - If $VIF_j > 10$, variable j is removed

Steps for the ABC Elastic Net regression



Csillery *et al.* (2010, *Trends Ecol Evol*)
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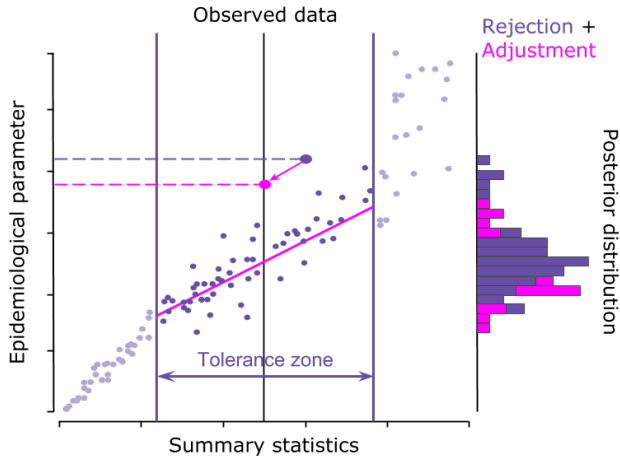
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2. ABC rejection step
 - Tolerance parameter = 10 %
 - Results in a first posterior distribution

Steps for the ABC Elastic Net regression

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Steps for the ABC Elastic Net regression



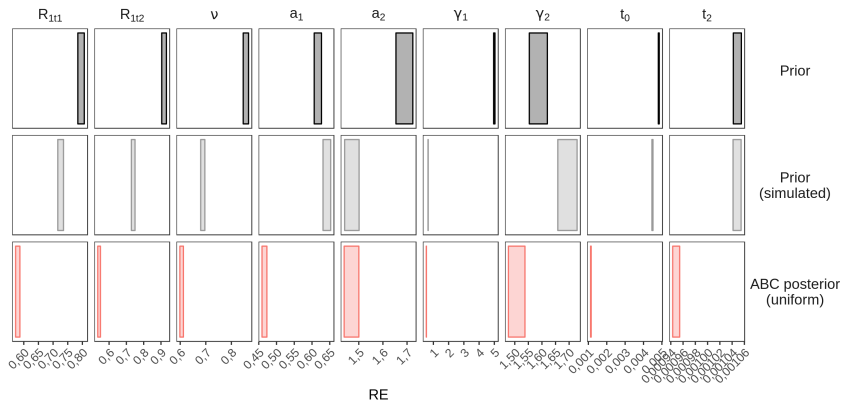
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Steps for the ABC Elastic Net regression

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2. ABC rejection step
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3. Weighting of the non rejected summary statistics
4. Adjust the posterior distribution using Elastic Net linear regression model
 - Combines LASSO regression penalty and Ridge regression penalty
 - LASSO: tends to set coefficients to zero
 - Ridge: shrinks coefficients
 - Results in a final posterior distribution

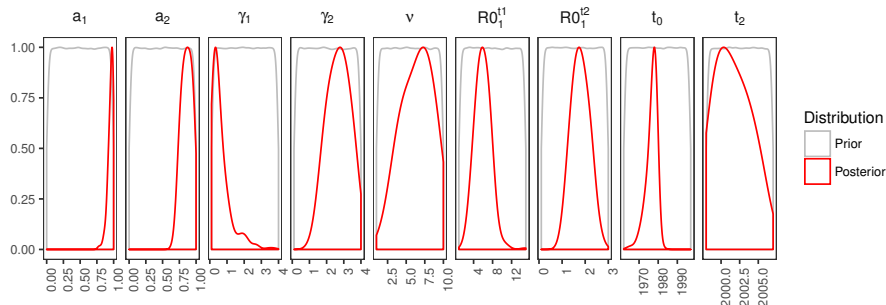
Cross-validation

- 100 replicates: 1 test data x 60.000 training data
- Inference error measurement using relative error (RE)



Given the phylogeny structure and the low number of sequences, our ABC regression method has the power to infer most parameters.

Posterior distributions



a_1	a_2	γ_1	γ_2	ν	R_0^{1,t_1}	R_0^{1,t_2}	R_0^{2,t_3}	t_0	t_2
0.96	0.86	0.6	2.7	6.5	5.9	1.79	2.35	1977	2002

$$1/\gamma_1 = 20 \text{ months}, 1/\gamma_2 = 4 \text{ months}$$

Conclusions

- New MSM hosts transmit more than classical IDU hosts.
- The R_0 for the "classical" and the "new" epidemics are respectively 1.80 and 2.35.
- The new epidemic is characterized by a shorter infectious period duration (4-5 months) than the classical epidemic (18 months).

Conclusions

- We developed new summary statistics to apply regression-ABC to a multi-hosts typed data to infer different R_0 .
- We inferred epidemiological parameters despite the low number of sequences.
- The cross-validation results show how robust our results are for most parameters.
- Perspectives: model comparison, correlations between parameters.

Thank you for your attention !

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- The new epidemic is characterized by a shorter infectious period duration (4-5 months) than the classical epidemic (18 months).
- We developed new summary statistics to apply regression-ABC to a multi-hosts typed data to infer different R_0 simultaneously.

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