Extensions to the pomp package and framework

Spencer J. Fox Qianying (Ruby) Lin Jesse Wheeler

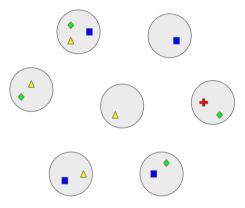
Package Extensions

Mathematically, POMP models are very versatile, and many of the ideas useful to the models described so far are also applicable in more general settings. This gives rise to a few useful extensions of the pomp package, which we try to describe here.

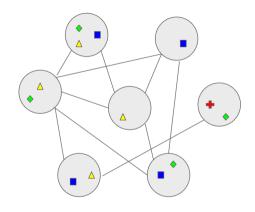
- Meta-population data and models
 - panelPomp
 - spatPomp
- ► Genomic / phylogenetic data
 - phylopomp

Each of these packages extend the pomp package to other useful cases.

Meta-population: Data and Models



panelPomp: data from related, independent systems.



spatPomp: data from a single interacting system.

Meta-population packages: Examples

Both packages require pomp to be installed, and the internal workings of both packages actually use many of the pomp functions. Thus, familiarity with pomp is a prerequisite.

- panelPomp:
 - We have *panel* or *longitudinal* data.
 - Several independent yet related systems. Information from all systems is useful, not just one location.
 - Simultaneously allows for features shared by each system as well as system unique features.
 - Example: endemic COVID-19 measured in 3 locations: New York, London, Atlanta.
- spatPomp:
 - ▶ The name comes from the idea that we have spatially explicit POMP models.
 - The data at multiple locations are thought of as location-specific measurements from the same system. That is, the dynamic systems underlying the data are connected.
 - Example: Modeling early stages of COVID-19 in China. Here, we might have measurements for many cities, but at first cases were only found in Wuhan.

PanelPomp: a collection of POMP models

A panelPomp model is really just a collection of pomp models. This is also how they are built:

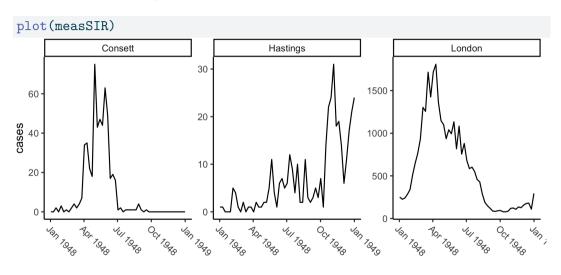
```
library(panelPomp)
mod1 \leftarrow pomp(..., params = c('p1' = 0.1, 'p2' = 1.2, 'p3' = 0.9))
mod2 \leftarrow pomp(..., params = c('p1' = 0.1, 'p2' = 1.1, 'p3' = 0.6))
mod3 \leftarrow pomp(..., params = c('p1' = 0.1, 'p2' = 1.5, 'p3' = 0.75))
ppomp <- panelPomp(</pre>
  object = list(mod1, mod2, mod3),
  shared = c("p1" = 0.1), # Shared-value parameters
  specific = c("p2", "p3") # Unit-Specific parameters
```

Measles Example

We can build off of the measles example by looking at UK measles from multiple cities, and building an SEIR model for the data. There is built in models and data in panelPomp

```
measSIR <- panelMeasles(
  units = c("Consett", "London", "Hastings"),
  first_year = 1948,
  last_year = 1948
)</pre>
```

Measles Example: Figure



panelPomp: parameters

Parameter names in panelPomp have the following conventions. If the parameter is shared, it just is called by the name. If it is unit-specific, the name of the unit follows the name of the parameter: cunit

coef(measSIR)

		0001 (1110000111)
gamma[Consett]	sigma[Consett]	mu
1.72e+02	4.26e+01	2.00e-02
amplitude[Consett]	RO[Consett]	rho[Consett]
2.00e-01	3.59e+01	6.50e-01
<pre>cohort[Consett]</pre>	<pre>iota[Consett]</pre>	alpha[Consett]
3.10e-01	7.31e-02	1.01e+00
E_0[Consett]	S_0[Consett]	<pre>psi[Consett]</pre>
1.83e-05	3.22e-02	4.06e-01
${\tt sigmaSE[Consett]}$	R_0[Consett]	I_0[Consett]
7.12e-02	9.68e-01	1.97e-05
rho[Hastings]	${\tt gamma[Hastings]}$	sigma[Hastings]
6.95e-01	7.41e+01	5.63e+01

panelPomp: shared vs unit-specific

A key feature of panelPomp objects is which parameters are shared, which are unit-specific. If parameters are shared, that means they have the same value for all units. We can inspect and modify which parameters are which using the functions shared and specific:

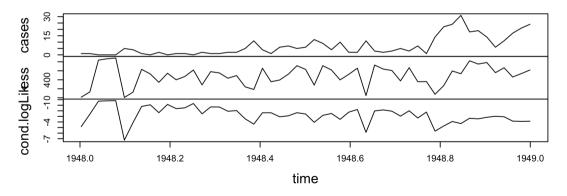
```
shared(measSIR)
  mu
0.02
shared(measSIR) <- c(shared(measSIR), 'alpha' = 1)
shared(measSIR)
  mu alpha
0.02 1.00</pre>
```

panelPomp Functions

panelPomp is the easiest extension: If you can build one pomp, you can build multiple into a single panelPomp object. Existing functions and algorithms are similar as well!

```
pfilter(measSIR, Np = 1000) |> plot(unit = 'Hastings')
```

Hastings



panelPomp: iterated filtering

We need a new algorithm for conducting maximum likelihood, either the panel iterated filter (PIF), or marginalized panel iterated filter (MPIF). We don't even need to change the code!

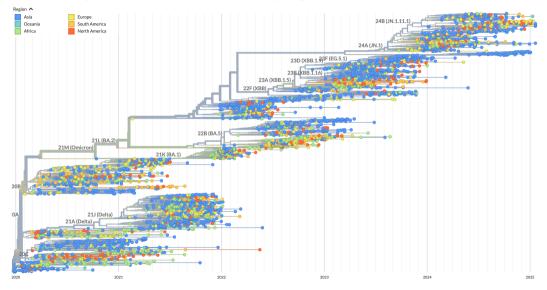
```
mif2Out <- mif2(
  measSIR, rw.sd = rw_sd(rho = 0.02, R_0 = 0.02),
  Nmif = 10, Np = 200, cooling.fraction.50 = 0.5,
  block = TRUE  # block = TRUE does MPIF, usually best + faster.
)</pre>
```

Phylodynamic package: phylopomp

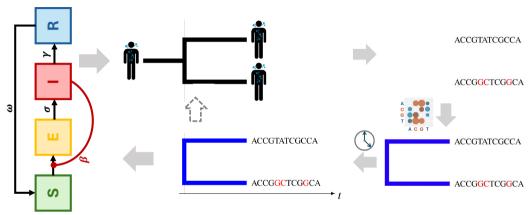
The spreading of infectious diseases can be considered as the mutation accumulation in genomic sequences in hosts and the transmission of such virus from host to host. Therefore, when the viral genomic samples are collected from hosts, one can traceback a partially observed ancestral history and reconstruct it as a tree, called genealogy or phylogeny.

With phylopomp, we can simulate these partially observed ancestral history from a board class of epidemiological models, and infer the transmission dynamics given a genealogy/phylogeny based on pomp.

phylopomp: genomic sequences & phylogenetic tree



phylopomp: phylogenetics & phylodynamics



The goal of phylopomp is to infer the partially observed epidemiological dynamics, formulated by compartmental models, using the genealogy/phylogeny, reconstructed from sampled genomic sequences.

phylopomp: installation

The github repo for phylopomp is on https://github.com/kingaa/phylopomp/. We can install the latest released version from github using devtools.

```
library(devtools)
install_github("kingaa/phylopomp@0.14.8.0")
```

To confirm the package is installed successfully,

```
packageVersion("phylopomp")
```

[1] '0.14.8.0'

library(phylopomp)

phylopomp: simulation and inference

- ➤ The first useful function of phylopomp is to build customized models and simulate genealogies from it. Pre-defined models are included: linear birth-death model (lbdp), moran model, SIR, SEIR, two-class SIR model with super-infection (si2r), two-strain SIR model (siir), etc.
- The second is to infer the models given genealogies using pomp. This function is currently under development, while a few models are available: 1bdp, moran, sir, and seir.

Example: simulation with pre-defined models in phylopomp I

```
set.seed(1234)
simulate(
  "SIR", time=2,
 Beta=2,gamma=1,psi=2,
  S0=1000, I0=5
  # update params
  simulate(
    time=5.
    Beta=5,gamma=2,psi=3
  ) -> model.sir
plot(model.sir)
```

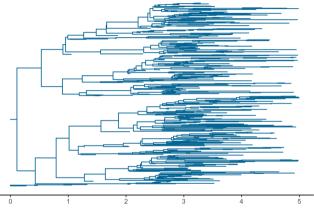


Figure 1: Simulated genealogy from an SIR model.

Example: simulation with pre-defined models in phylopomp II

```
set.seed(1234)
simulate(
  "SEIR",time=2,
  Beta=2, sigma=2,
  gamma=1,psi=2,
  S0=1000.I0=5
  simulate(
    time=5.
    Beta=5,gamma=2,
    psi=3
  ) -> model.seir
plot(model.seir)
```

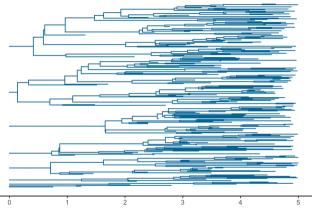


Figure 2: Simulated genealogy from an SEIR model.

Example: simulation with pre-defined models in phylopomp III

```
# show E/I transitions
model.seir |>
  plot(obscure=FALSE)
```

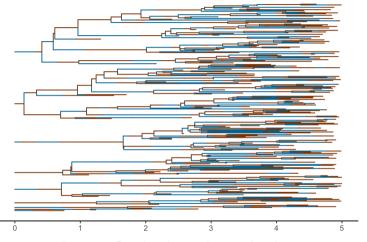


Figure 3: Simulated genealogy with colors.

Example: simulation with pre-defined models in phylopomp IV

```
model.seir |>
  lineages(
   obscure=FALSE
) |>
  plot()
```

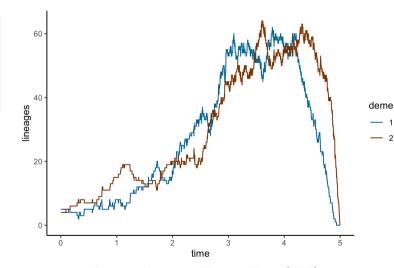


Figure 4: Lineages Through Time (LTT).

Example: inference with pre-defined models in phylopomp I

```
library(pomp)
set.seed (1234)
# simulate a geneal object from an SIR model
simulate("SIR", time=10, Beta=3, gamma=1, psi=2, omega=1, S0=100, I0=5) -> x
# build a pomp object
x |>
  sir pomp(
    Beta=3,gamma=1,psi=2,omega=1,
    S0=100, I0=5, R0=0
  ) -> po
po |> pfilter(Np=5000) -> pf
pf |> logLik()
[1] -326.5652
```

Example: inference with pre-defined models in phylopomp II

```
set.seed(1234)
simulate ("SEIRS", # simulate a genealogy from an SEIR model
  Beta=4, sigma=1, gamma=1, psi=1, omega=1,
  S0=100.E0=3.I0=5.R0=100. time=5
) -> G
G |>
  seirs pomp(
    Beta=4, sigma=1, gamma=1, psi=1, omega=1,
    S0=100, E0=3, I0=5, R0=100
  ) |> pfilter(Np=1000) |>
  replicate(n=20) |> concat() -> pf
pf |> logLik() |> logmeanexp(se=TRUE)
        est.
                     se
-168.856637 1.265223
```

License, acknowledgments, and links

- ▶ This lesson is prepared for the Simulation-based Inference for Epidemiological Dynamics module at the Summer Institute in Statistics and Modeling in Infectious Diseases, SISMID.
- The materials build on previous versions of this course and related courses.
- Licensed under the Creative Commons Attribution-NonCommercial license. Please share and remix non-commercially, mentioning its origin.
- Produced with R version 4.4.2 and pomp version 6.3.
- Compiled on 2025-07-23.

Back to Lesson R code for this lesson