Estimating Epidemiological Parameters of an Ebola Outbreak using BEAST2 Tutorial by <u>Tracy Heath</u> and <u>Tanja Stadler</u> This tutorial uses the sequence data from 72 Ebola patients in Sierra Leone. The viral sequences were first presented in Gire et al. (Science 2014). These data were re-analyzed under complex birth-death processes

to estimate key epidemiological parameters by Stadler et al. (PLoS Currents Outbreaks 2014). In this exercise, we will perform a simplified analysis similar to one conduced by Stadler et al. (2014). We will use the birth-death process with serial sampling and piecewise shifts in rates (the "BD" model in the

Stadler et al. 2014 study). Please refer to Stadler et al. (2014) and Stadler et al. (2013) for detailed descriptions of these models and methods.

For more information about divergence-time estimation in general and BEAST v2.2, please refer the resources and tutorials on: http://phyloworks.org/workshops/divtime.html. In particular see the detailed tutorial on estimating speciation times using extant and fossil data and the links and references therein: http://treethinkers.org/tutorials/divergence-time-estimation-using-beast/.

- or treated, their behavior changes, or they die. [called "becomeUninfectiousRate" in BEAST2] • s is the probability of sampling an individual upon becoming non-infectious. [called "samplingProportion" in BEAST2] • λ is the rate of transmission (birth rate). [called "birth" in BEAST2] μ is the viral lineage death rate. [called "death" in BEAST2] ψ is the rate each individual is sampled. [called "sampling" in BEAST2]
- In the analyses, we will estimate R, δ , and s. The remaining parameters listed above can be calculated if
- In this exercise, we also assume that there are *l* intervals, evenly spaced, over the tree. These intervals are

the probabilistic graphical model depicting the conditional dependence structure of all of the parameters in our skyline birth-death model. It is also possible to consider a model the other parameters (δ , s, μ , Ψ) also change over time as in Stadler et al. (2013), however, we will not consider that more complex model in this exercise.

rate of becoming non-infectious effective reproductive number

time tree

Probabilistic graphical model of the skyline birth-death process. The parameters R, δ , s, λ , μ , and Ψ are defined above, as is l. The

origin time x_0 , is the start of the epidemic. This graph shows that the time tree T is conditionally dependent on all of the parameters

death rate

origin time

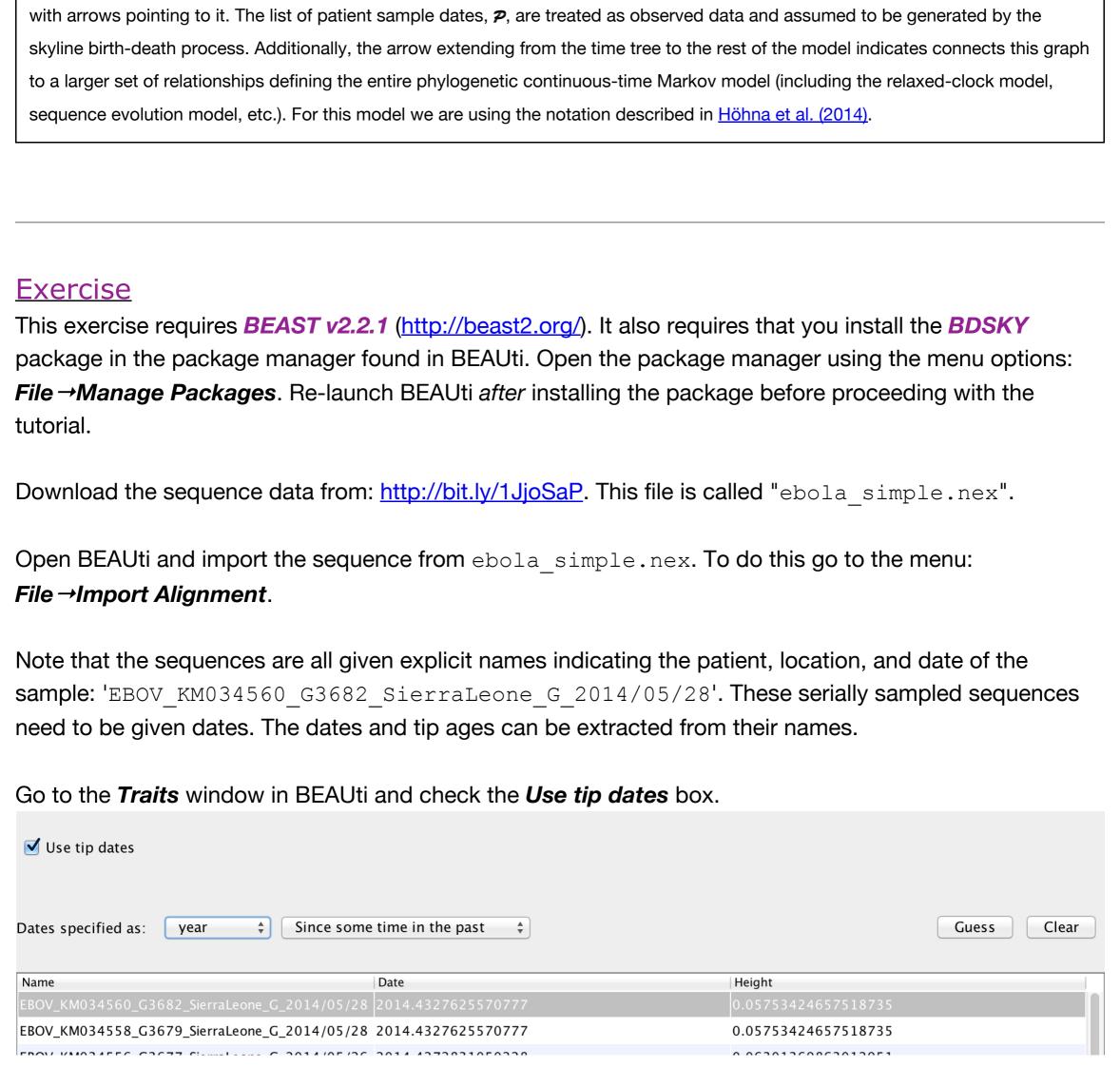
patient sample dates

 x_0

sampling rate

transmission rate

 $i \in \{1, \dots, l\}$



rate.

Origin

М

S

Offset

Alpha

4.592E-4.

Mean

Sigma

Offset

Alpha

Beta

Offset

clockRate.c:ebola_simple

samplingProportion.t:ebola_s...

Go to the **Priors** window.

Tree.t:ebola simple

Become Uninfectious Rate

deviation of 1.25.

R.t:ebola_simple

Mean In Real Space

becomeUninfectiousRate.t:eb...

Change the tree prior to **Birth Death Skyline Serial**

100.0

2.0

1.0

0.0

1.25

0.0

0.5

61.0

Birth Death Skyline Serial

Log Normal

Gamma

Normal

0.001984

4.592E-4

0.0

10.0

6.0

0.0

number changed over the course of the epidemic.

Change the *Chain Length* to 20,000,000.

This file should now run in BEAST.

500

sequences.

Summary Statistic

0.08

0.07

0.06

0.05

0.04

0.03

0.02

0.01

0.0

in the **Estimates** panel.

R.1

mean 5.074

stdev 3.1577

variance 9.9709

median 4.7697

mode n/a

95% HPD Interval [0.0289, 10....

stderr of mean 0.2033

geometric mean 3.7514

auto-correlation time (ACT) 74637.8845

effective sample size (ESS) 241.1912

over time. We want to change this.

character and click **OK**.

estimate 0.01 Sampling Proportion Set the prior on the effective reproductive rate (R) to a lognormal with a log-mean of 0.0 and standard

initial = [2.0] [0.0,Infinity]

initial = [1.0] [0.0,Infinity]

estimate 🥖

🔲 estimate 🥖

estimate 🥖

estimate 🥖

estimate 🥖

estimate 🥖

estimate 🥖

prior over R.t:ebola_simple

2.50

5.00

7.50

2.5% Quantile 5% Quantile

95% Quantile

97.5% Ouantile

prior over becomeUninfectiousRate.t:ebola_simple

10.0

7.82

Median 1.00

0.00 0.000500 0.00100 0.00150 0.00200 0.00250 0.00300 0.00350 (

2.5% Quantile 5% Quantile

95% Quantile

97.5% Quantile

Median

0.00108

0.00123

0.00198

0.00274

0.700

0.500-

0.400-

0.300 0.200-

0.100 0.00-

0.00

0.0400

0.0350 0.0300

substitution rate of partition c:ebola_simple

800 700-

600-500-400-

300-200-100-

initial = [0.01] [0.0,1.0] prior on sampling proprtion t:ebola_simple

2.50-2.00-

1.50-1.00-0.500 0.00-

0.300

0.400

0.500 2.5% Quantile 5% Quantile

-

For this analysis, in the Clock Model window we will leave the default Strict Clock and estimate the clock

Beta estimate 🥖 0.0250-0.0200-0.0 Offset 0.0150-0.0100-0.00500-0.00-150 200 250 100 0.00 2.5% Quantile 0.0300 5% Quantile Median 13.9 95% Quantile 117 97.5% Ouantile

Set the prior on the clock rate to a normal distribution with a mean of 0.001984 and a standard deviation of

initial = [1.0] $[-\infty, \infty]$

\$

Specify a Beta prior on the sampling proportion with parameters $\alpha=10.0$ and $\beta=6.0$.

Now go to the Initialization menu. To reveal this, you have to go to *View →Show Initialization panel*.

By default, the birth-death skyline model assumes that there are 9 shifts in the R, δ , and s parameters

Change the *Dimension* to "1" for the **samplingProportion** and **becomeUninfectiousRate**. Here, we are

assuming that these parameters remain constant over time, even though the effective reproductive

Set the sampling frequency (*Log Every*) to 2,000 iterations for both the *tracelog* and *treelog*.

Save the XML file by going to *File →Save As*. (Perhaps name the file "ebola_simple.xml".)

For **R**, set the **Dimension** to "3". This means that we are assuming a model where the effective

reproductive number changed two times after the start of the epidemic.

Go to the **MCMC** panel to set the sampling frequency and chain length.

Specify a Gamma prior on the rate of becoming non-infectious with a shape of 0.5 and a scale of 61.

R.1 R.2 R.3

death process discussed here, one must alter the XML file directly. Summarize the posterior sample of trees using TreeAnnotator. 000 TreeAnnotator v2.2.1 Burnin percentage: 20 Posterior probability limit: 0.0 Maximum clade credibility tree Target tree type: Node heights: Common Ancestor heights Choose File... not selected Target Tree File: Choose File... Input Tree File: ebola simple.trees Output File: ebola_simple.tan.tre Choose File... Quit Run The maximum clade credibility tree has very low support for these data.

12.5 10 7.5 2.5 The estimates of the reproductive number and origin times differ from those reported in the Stadler et al. (2014) study. This is because we are using slightly different priors and a reduced dataset. It isn't currently possible to replicate the model used in their study by simply setting up the analysis in BEAUti. This more complex model, that includes variable sampling through time and other extensions of the piecewise birth-

When the run terminates, open the log file in *Tracer* and view the summaries of each parameter. Look at the estimate of the *origin* time. The origin is the time of infection of the first person in the outbreak (Stadler et al. 2014). Below is a histogram of the origin time from an analysis of these data. The mean estimated time is 0.0902. We can compute the origin time in terms of the number of days before the last sample since we know that our time is in units of years. Thus, the origin time is 32.9 days before the most recently sampled sequence: $\frac{0.0902}{1} = 32.9$ The last sequences were sampled on 18 June 2014, which is the 169th day of 2014. Thus, the mean origin time corresponds to 17 May 2014 (the 137th day of 2014), with a 95% highest posterior density interval of 26 April to 25 May. **Summary Statistic** mean 0.0902 stderr of mean 2.5121E-3 stdev 0.0324 variance 1.0518E-3 median 0.0799 mode n/a geometric mean 0.0867 95% HPD Interval [0.0666, 0.1454] auto-correlation time (ACT) 1.0802E5 effective sample size (ESS) 166.6509 2500

2000 1500° Frequency

0.2

R.2

1.3904

0.0709

1.1833

1.4003

1.0726

n/a

0.988

[0.0264, 3.7...

64603.3388

278.6543

origin

Calculate the mean and 95% HPD interval dates for the *TreeHeight* parameter using the formulas

given above. This parameter is the date of the root or most-recent-common-ancestor of all sampled

You can visualize the change in the effective reproductive number over time by selecting R.1, R.2, and R.3

R.3

3.7277

0.0674

1.4171

2.0083 3.4404

[1.5986, 6.5...

40757.2034

441.6888

n/a 3.5005 0.3

0.4

0.5

1.0 🔲 estimate 🥖

With the dates specified as *year* and *Since some time in the past*, click on the button called *Guess*.

BEAUti will extract the dates from the sequence names, given that they are provided using a specific text

pattern. Here, the dates all follow the last "_" character. Indicate that you want to use everything after this

after last

*

🗹 estimate 🧷

🔲 estimate 🥖

🗹 estimate 🥖

\$

estimate

estimate

estimate

15.0

17.5

You can see now that each sequence is given a date that is a value relative to year 0. Thus, the sequence sampled on 28 May 2014 is 2014.4327625570777. The tip heights are computed relative to the most recent sample, which has a height of 0.0. The units are in years, thus one day is a difference in tip height of 1/365 = 0.002739726. Go to the Site Model window and change the sequence model to HKY+G. Substitution Rate 4 Gamma Category Count Shape 1.0 0.0 Proportion Invariant HKY 2.0 Kappa Frequencies Estimated

use everything

we have the first three because they are defined: $R = \frac{\lambda}{\mu + \psi}, \quad \delta = \mu + \psi, \quad s = \frac{\psi}{\mu + \psi}$ Thus we can compute λ , μ , and Ψ given: $\lambda = R\delta, \ \mu = \delta - s\delta, \ \psi = s\delta$ delineated by shifts in the value of R and because λ is determined by R, λ also changes over time. Below is probability of sampling