

1 **Submission type:** Resource article

2

3 **PhyloSTemS: a new graphical tool to investigate temporal signal of**
4 **heterochronous sequences at various evolutionary scales**

5

6 Anna Doizy¹², AmauryPrin¹, Guillaume Cornu³, Frederic Chiroleu¹ and Adrien Rieux^{1*}

7 ¹ CIRAD, UMR PVBMT, 97410 St Pierre, La Réunion, France.

8 ² DoAna - statistiques Réunion, F-97480 Saint-Joseph, Reunion island, France

9 ³ CIRAD, Univ Montpellier, UR Forests and Societies, Montpellier, France

10 *Author to whom correspondence should be addressed: adrien.rioux@cirad.fr

11

12

13

14 **Keywords:** Measurably-evolving populations (MEPs), heterochronous sequence dataset,
15 temporal signal, root-to-tip regression, phylogenetic tip-dating, R shiny app

16 **Running title:** Phylogenetic scaling of temporal signal

17 **Conflict of interest:** None declared.

18

19 **Abstract**

20 Molecular tip-dating of phylogenetic trees is a growing discipline that uses DNA sequences
21 sampled at different points in time to co-estimate the timing of evolutionary events with
22 rates of molecular evolution. Such inferences should only be performed when there is
23 sufficient temporal signal within the analysed dataset. Hence, it is important for researchers
24 to be able to test their dataset for the amount and consistency of temporal signal prior to
25 any tip-dating inference. For this purpose, the most popular method considered to-date has
26 been the “root-to-tip regression” which consist in fitting a linear regression of the number of
27 substitutions accumulated from the root to the tips of a phylogenetic tree as a function of
28 sampling times. The main limitation of the regression method, in its current implementation,
29 relies in the fact that the temporal signal can only be tested at the whole tree evolutionary
30 scale. To fill this methodological gap, we introduce PhyloSTemS, a new graphical and user-
31 friendly tool developed to investigate temporal signal at every evolutionary scale of a
32 phylogenetic tree. PhyloSTemS allows detecting without *a priori* whether any subset of a
33 tree would contain sufficient temporal signal for tip-based inference to be performed. We
34 provide a “how to” guide by running PhyloSTemS on empirical datasets and supply guidance
35 for results interpretation. PhyloSTemS is freely available at [https://pvtbmt-](https://pvtbmt-apps.cirad.fr/apps/phylostems)
36 [apps.cirad.fr/apps/phylostems](https://pvtbmt-apps.cirad.fr/apps/phylostems).

37

38

39

40

41 Introduction

42 “Tip-dating” of phylogenetic trees is a popular and powerful type of genetic analysis aiming
43 to make use of sequence data isolated at different points in time (i.e., heterochronous
44 datasets) to co-estimate the timing of evolutionary events with rates of molecular evolution
45 (Rieux & Balloux, 2016). Tip-dating requires working on measurably evolving populations
46 (MEPs) which consist in datasets displaying detectable amounts of *de novo* nucleotide
47 changes among the DNA sequences sampled at different timepoints (Drummond, Pybus,
48 Rambaut, Forrest, & Rodrigo, 2003). Our ability to capture measurable amount of
49 evolutionary change from sequence data is a factor of various parameters including the
50 evolutionary rate per site per unit of time (μ), the width of the sampling interval (t), the
51 number of sites in the sequences (L) and the time to the Most Recent Common Ancestor
52 (MRCA) of all sequences (T_{MRCA}). Originally, only fast-evolving organisms such as RNA viruses
53 were classifiable as MEPs but the recent rise in our ability to sequence DNA at high
54 throughput from both modern and ancient material has led to a massive increase in both
55 sequence length (L) and the timespan covered by the sequences (t), hence opening up the
56 field of tip-dating to a variety of additional organisms (Biek, Pybus, Lloyd-Smith, & Didelot,
57 2015).

58 Phylogenetic inferences performed on such time-structured sequence data represent a
59 powerful tool for hypothesis testing (Rieux & Balloux, 2016). They have notably been critical
60 for *i*) dating key events in human evolutionary history (Fu et al., 2013; Rieux et al., 2014), *ii*)
61 improving our understanding of various important pathogens emergence, spread and
62 evolution (Bos et al., 2014; Eldholm et al., 2015; Faria et al., 2014; O’Hanlon et al., 2018;

63 Rambaut, 2020; Vanhove et al., 2019), *iii*) investigating the relative impacts of climatic and
64 anthropogenic factors on the widespread extinctions of large mammals (Shapiro et al., 2004;
65 Stiller et al., 2010), *iv*) providing meaningful information about pathogens host species jumps
66 (Weinert et al., 2012) and *v*) estimating unknown molecular sequence's ages in various
67 organisms (Shapiro et al., 2011).

68 Inferences from tip-calibrated phylogenetic trees should only be performed when there is
69 sufficient temporal signal within the analysed dataset (Drummond, Pybus, Rambaut, et al.,
70 2003; Duchêne, Duchêne, Holmes, & Ho, 2015; Murray et al., 2016; Rieux & Balloux, 2016).
71 This will for instance not be the case if the sampling period is too short for sufficient
72 evolutionary changes to be measured, if evolutionary rates are too variable or if some
73 samples have incorrectly been dated (Rambaut, Lam, Carvalho, & Pybus, 2016). As such it is
74 important for researchers to be able to test their dataset for the amount and consistency of
75 temporal signal prior to any tip-dating inference. For this purpose, the most popular method
76 considered to-date has been the "root-to-tip regression" which consist in fitting a linear
77 regression of the number of substitutions accumulated from the root to the tips of a
78 phylogenetic tree as a function of sampling times (Buonagurio et al., 1986; Drummond,
79 Pybus, & Rambaut, 2003; Korber et al., 2000; Shankarappa et al., 1999). If sampling dates are
80 sufficiently different, then more recently sampled sequences should have undergone
81 substantially more evolutionary change than earlier sampled sequences, which would result
82 in a positive correlation. This method has often been used as a diagnostic of data quality and
83 of the reliability rate estimates, where the slope coefficient corresponds to the substitution
84 rate under the assumption of a strict molecular clock, the X-intercept is an estimate of the
85 date of the root of the tree and R^2 indicates the degree to which sequence evolution has

86 been clocklike. However, the root-to-tip regression method is not statistically suitable for
87 proper hypothesis testing because the individual data points are not independently
88 distributed, and are instead partially correlated due to their phylogenetic shared ancestry
89 (Drummond, Pybus, & Rambaut, 2003). To overcome this limitation, Navascues et al. (2010)
90 suggested a non-parametric approach using permutations to test whether the correlation is
91 stronger than expected if the sampling dates were randomly assigned. Linear regression is a
92 crude method but other phylogenetic approaches such as the date-randomization test
93 (Duchêne et al., 2015; Duffy & Holmes, 2009; Murray et al., 2016; Ramsden, Melo,
94 Figueiredo, Holmes, & Zanotto, 2008) or model selection/comparison (Duchene et al., 2019;
95 Murray et al., 2016; Rambaut, 2000), although way more computationally intensive, have
96 recently been introduced and shown to be more robust tests for temporal signal detection
97 and characterization.

98 Despite its statistical pitfalls, the regression method remains a very helpful exploration tool
99 to quickly assess the extent of temporal signal within a dataset. It only requires a rooted
100 molecular phylogeny (whose branch lengths represent genetic distance) estimated from
101 heterochronous (dated) sequences and runs instantaneously. The regression method has
102 been implemented in the popular and interactive graphical program TempEst (Rambaut et
103 al., 2016), formerly known as Path-O-Gen. The main limitation of the regression method in
104 its current implementation relies in the fact that the temporal signal can only be tested at
105 the whole dataset (tree) evolutionary scale. However, although a significant positive
106 correlation would indicate the presence of detectable amounts of *de novo* mutations within
107 a tree timescale, a non-positive (or a statistically non-significant) correlation does not
108 necessarily mean that no temporal signal exists at a reduced timescale, as illustrated in Fig 1.

109 To fill this methodological gap, we introduce PhyloSTemS, a new graphical and user-friendly
110 tool developed to investigate temporal signal at every evolutionary scales of a phylogenetic
111 tree. PhyloSTemS allows detecting without *a priori* whether any subset of a tree would
112 contain sufficient temporal signal for tip-based inference to be performed. We provide a
113 “how to” guide by running PhyloSTemS on empirical datasets and supply insights on
114 interpreting the outputs.

115 **Materials and Methods**

116 The program PhyloSTemS (Phylogenetic Scaling of Temporal Signal) is an open source,
117 graphical Shiny based R application (Chang, Cheng, Allaire, Xie, & McPherson, 2018; R Core
118 Development Team, 2020) built for exploring temporal signal at various scales of a
119 phylogenetic tree. Shiny is an R package that makes it easy to build interactive web
120 applications from R (<https://shiny.rstudio.com/>). PhyloSTemS can be either used online at
121 <https://pvbmt-apps.cirad.fr/apps/phylostems/> or executed locally by downloading its source
122 code from <https://gitlab.com/cirad-apps/PhyloSTemS>. A schematic representation of
123 PhyloSTemS workflow is presented in Figure 2.

124 As input, PhyloSTemS requires (i) a phylogenetic tree in computer-readable Nexus or Newick
125 format with branch lengths scaled as genetic distances only, such as the ones computed
126 using maximum likelihood approaches (e.g. Guindon et al., 2010; Minh et al., 2020;
127 Stamatakis, 2014). In its current implementation, the online version of PhyloSTemS allows
128 uploading trees with 1500 sequences at maximum. Larger trees will need to be processed
129 locally by sourcing the gitlab version. (ii) Prior to be loaded in PhyloSTemS, the tree needs to
130 be rooted, either at a position chosen by the user (with an outgroup) or at a most

131 compatible location with the assumption of a strict molecular clock (using for instance the
132 rtt function from the ape R package (Paradis & Schliep, 2019)). When possible, we advise to
133 use outgroup-rooted trees. Finally, *(iii)* sampling/isolation dates needs to be known for each
134 sequences and specified as calendar numeric years (e.g 1986.4 or 2017) within tip labels.
135 Before-Christ (B.C) dates, sometimes required to handle sequences generated from ancient
136 DNA data can be specified using negative values (e.g. - 400.5). Note that since missing dates
137 are not allowed, sequences with unknown sampling years needs to be pruned out from the
138 tree (using for instance the drop.tip function from the ape R package) prior to be uploaded
139 in PhyloSTemS.

140 When a tree has correctly been loaded in PhyloSTemS, a distribution of sampling dates is
141 plotted within the “upload” panel allowing for a visual check of sequences temporal width.
142 At this stage, the phylogenetic tree has been loaded using the ape R package and root-to-tip
143 distances for all sequences are recorded. Temporal signal is hence tested at every node of
144 the input tree (including its root) meeting the following conditions required to perform a
145 linear regression: *i)* the node must be the parent of at least $n=3$ tips, *ii)* there should be at
146 least $n=3$ distinct combination of root-to-tip distances and sampling dates and *iii)* there
147 should be at least $n=2$ different sampling dates. At each of such nodes, linear regression
148 between sampling dates and root-to-tip distances is performed and the following
149 parameters: (1) p-value, (2) slope, (3) adjusted R^2 , and (4) intercept with the x-axis values are
150 recorded.

151 PhyloSTemS’s main results are provided within the “Temporal signal” panel. First an
152 annotated phylogenetic tree is interactively plotted by sourcing both ggtree and plotly

153 packages (Sievert, 2020; Yu, Smith, Zhu, Guan, & Lam, 2017). On this tree, nodes with
154 temporal signal, *i.e.* nodes at which root-to-tip linear regression yielded a statistically
155 significant and positive slope, are highlighted with colours scaling to R^2 value. The default
156 threshold for the linear regression p-value has been fixed to 0.05 but the user can
157 interactively modify it using a slider bar, which enable easy investigation of nodes with
158 borderline significant trends. A table summarizing the nodes with temporal signal is also
159 displayed along with respective number of descending sequences, p-value, slope and
160 adjusted R^2 values. Most importantly, PhyloSTemS allow the user to visualize the root-to-tip
161 regressions at any chosen node of interest. To do so, the user simply needs to click on a
162 node, and the associated root-to-tip regression will be displayed. Both the tree and the root-
163 to-tip regression plots are linked, so that data points (or tree tips) selected in one plot will
164 automatically be highlighted on the other one. This enables easy investigation of outliers and
165 sequences or clades of interest.

166 Finally, when temporal signal is found at the within-tree scale, PhyloSTemS's "Make new
167 FASTA" panel allows generating a new subset sequence FASTA file that only include the
168 variant sites for the descending tips of a node of interest, a dataset suitable for further tip-
169 dating inferences.

170 In the following, we use two previously published empirical datasets to illustrate how
171 PhyloSTemS allows users exploring temporal signal at various evolutionary scales within
172 phylogenetic trees. For both datasets, we downloaded rooted-ML tree files built from non-
173 recombining genomic sequences from their original publications. The first dataset contains
174 45 strains of *Xyllela fastidiosa* (hereafter *Xf*) sampled worldwide between 1983 and 2016

175 (Vanhove et al., 2019). *Xf* is a bacterial crop pathogen of global importance, currently
176 threatening agriculture in various European countries (Sicard et al., 2018). The second
177 dataset comprises 98 hantaviruses isolates sampled from bank voles in Belgium between
178 1984 and 2016 (Laenen et al., 2019). Hantaviruses are important zoonotic viral pathogens
179 that can cause hemorrhagic fever with renal syndrome and pulmonary syndrome, potentially
180 life-threatening diseases in humans (Maes, Clement, Gavrillovskaya, & Van Ranst, 2004).

181 **Results**

182 *Xf* dataset

183 We first loaded the *Xf* rooted tree within PhyloSTemS's "upload" panel (see Fig. 3). Looking
184 at the plot of the sampling dates distribution, one can perform a quick visual check of
185 sequences temporal width (here 1983-2016) to validate the data importation process.
186 Moving to the "Temporal signal" panel, PhyloSTemS displays the *Xf* phylogenetic tree on
187 which the structuration by the four subspecies: *ssp. pauca*, *multiplex*, *morus*, *fastidiosa* can
188 be easily distinguished (see Fig. 4A). Visual inspection of the *Xf* tree in PhyloSTemS
189 demonstrated a lack of strong and deep temporal signal as neither the root nor the MRCA of
190 each subspecies displayed any significant correlation between root-to-tip distances and
191 sampling ages, as highlighted by the absence of annotations at those nodes. PhyloSTemS
192 detected only one internal node (node 82) associated with temporal signal within the *Xf*
193 tree. This node is the MRCA of a small clade containing 9 samples within the *Xf pauca ssp*
194 clade. When clicking on this node, PhyloSTemS displays the associated root-to-tip regression
195 plot and parameters ($R^2 = 0.38$, slope = $6.9E-7$, P-val = 0.045, see Fig.4 B). According to
196 PhyloSTemS's results, this small clade (N=9 sample) is the only evolutionary scale suitable for

197 phylogenetic tip-based inferences in BEAST or other programs within the *Xf* dataset. To do
198 so, the “Make new FASTA” panel allows generating a new sequence file that only include the
199 variant sites for the 9 *Xf* samples within the clade with detected temporal signal.

200 Hantaviruses dataset

201 Visual inspection of the Hantaviruses tree in PhyloSTemS demonstrated heterogenous
202 temporal signal amongst clades, here referring to three geographical sampling areas namely
203 Ardennes, Campine and Sonian Forest (Fig 5.A). PhyloSTemS revealed the absence of
204 temporal at the whole tree scale since no temporal signal was found at the tree root. The
205 Sonian Forest clade was also associated with lack of temporal signal. Temporal signal was
206 observed at the MRCA of the Campine and Ardennes clades as well as within the Ardennes
207 clade, as represented by the several highlighted nodes on the tree. A table listing all the
208 nodes associated with temporal signal along with their associated statistics is given in Fig
209 5.B. When plotting the regression at the MRCA of the Campine and Ardennes clades,
210 PhyloSTemS allows visually identifying outlier samples that are significantly deviating from
211 the root-to-tip regression line (Fig 5.C). Here, all outliers fell within the Campine clade,
212 suggesting that phylogenetic tip-based inferences should not be performed on both the
213 Campine and Ardennes clades simultaneously. Possible causes for such outliers are multiple
214 and will be argued in the discussion section.

215 **Discussion**

216 We introduce PhyloSTemS, a new graphical and user-friendly tool developed to investigate
217 temporal signal within phylogenetic trees using the root-to-tip regression method. Previous
218 implementations of this method, such as for instance in the popular and interactive

219 graphical program TempEst (Rambaut et al., 2016) were designed to test temporal signal at
220 the whole tree scale (i.e. at its root). Investigating temporal signal at smaller evolutionary
221 scales was previously doable, but this task required the user to *i) a priori* decide at which
222 scale (i.e. on which samples) performing the test and *ii)* manually splitting or reconstructing
223 the tree for every of such scales. The main improvement of PhyloSTemS is to allow
224 detecting, in a single step and without *a priori*, any evolutionary scale at which temporal
225 signal may exist within a phylogenetic tree.

226 Exploring the degree of temporal signal in heterochronous sequences datasets before
227 proceeding to inference using formal molecular clock models is a crucial task (Rieux &
228 Balloux, 2016). As illustrated by the two empirical datasets analyzed in this study, temporal
229 signal may sometimes be heterogeneous within a tree with substantial differences between
230 clades. In such cases, we hope that PhyloSTemS will help researchers detecting the most
231 appropriate scales, if any, at which thorough tip-based inferences may be performed.
232 However, because of the statistical pitfalls associated with the root-to-tip regression method
233 (Rambaut, 2000; Rambaut et al., 2016), PhyloSTemS should rather be seen as a fast, visual
234 and qualitative data exploration tool for temporal signal detection but should not be used to
235 test hypotheses or undertake statistical model selection. Once temporal signal has been
236 detected in PhyloSTemS, we advise users to make use of other available methods such as
237 non-parametric permutations (Navascués et al., 2010), date-randomization test (Duchêne et
238 al., 2015; Duffy & Holmes, 2009; Murray et al., 2016; Ramsden et al., 2008) or model
239 selection/comparison (Duchene et al., 2019; Murray et al., 2016) to validate the existence of
240 measurably evolving populations in their datasets.

241 Finally, PhyloSTemS can also help identifying outliers or groups of samples that substantially
242 differ from the root-to-tip regression line and may require careful handling to avoid bias
243 during phylogenetic inferences. First, as illustrated by the analyse of the Hantaviruses
244 dataset, different clades or populations in a tree may be characterized by positive but
245 contrasted root-to-tip regression patterns that might arise from sampling bias or differences
246 in life-history traits between clades (e.g. environmental factors, population density,
247 evolutionary rates or epidemiological parameters). In such a case, it is suggested to perform
248 independent phylogenetic inferences on each clade/population (Laenen et al., 2019). In
249 other cases, outlier sequences whose sampling date is incongruent with their genetic
250 divergence and phylogenetic position can be spotted from the regression plot (Rambaut et
251 al., 2016). Such anomalies can reflect a problem with *i)* the sequence itself (e.g. low quality,
252 sequencing/assembly/alignment errors, recombination or hypermutation) or *ii)* the sampling
253 date(s) (e.g. mislabelling or biological contamination). Should the case of such outlier
254 sequences arise, those samples should be excluded from subsequent phylogenetic
255 inferences.

256 Considering the impressive increase in availability and use of heterochronous datasets, we
257 hope the functionality provided by PhyloSTemS will help users to perform thorough tip-
258 dating inferences. PhyloSTemS is a dynamic application by nature. New functions will be
259 added as new needs arise.

260

261

262

263 **Acknowledgements**

264 This work was financially supported by l'Agence Nationale pour la Recherche (JCJC
265 MUSEOBACT contrat ANR-17-CE35-0009-01), the European Regional Development Fund
266 (ERDF contract GURDT I2016-1731-0006632), Région Réunion and the French Agropolis
267 Foundation (Labex Agro – Montpellier, E-SPACE project number 1504-004). We are grateful
268 to S. Falala for his advices on building Shiny apps and CIRAD for providing hosting of the
269 application server. We thank R. Almeida, M. Vanhove & B. Vrancken for providing access to
270 the empirical datasets analyzed in this study. We also thank L. Van Dorp, P. Campos, C.G.
271 Crego, E. Conte, F. Balloux & D. Richard for testing previous versions of the app on their own
272 datasets.

273

274

275

276

277

278

279

280

281

282

283

284 **References**

- 285 Biek, R., Pybus, O. G., Lloyd-Smith, J. O., & Didelot, X. (2015, June 1). Measurably evolving
286 pathogens in the genomic era. *Trends in Ecology and Evolution*, Vol. 30, pp. 306–313.
287 doi: 10.1016/j.tree.2015.03.009
- 288 Bos, K. I., Harkins, K. M., Herbig, A., Coscolla, M., Weber, N., Comas, I., ... Krause, J. (2014).
289 Pre-Columbian mycobacterial genomes reveal seals as a source of New World human
290 tuberculosis. *Nature*, 514(7253), 494–497. doi: 10.1038/nature13591
- 291 Buonagurio, D. A., Nakada, S., Parvin, J. D., Krystal, M., Palese, P., & Fitch, W. M. (1986).
292 Evolution of human influenza A viruses over 50 years: Rapid, uniform rate of change in
293 NS gene. *Science*, 232(4753), 980–982. doi: 10.1126/science.2939560
- 294 Chang, W., Cheng, J., Allaire, J., Xie, Y., & McPherson, J. (2018). shiny: Web Application
295 Framework for R. <https://CRAN.R-Project.Org/Package=shiny>. Retrieved from
296 <https://cran.r-project.org/package=shiny>
- 297 Drummond, A., Pybus, O. G., & Rambaut, A. (2003). Inference of Viral Evolutionary Rates
298 from Molecular Sequences. *Advances in Parasitology*, Vol. 54, pp. 331–358. doi:
299 10.1016/S0065-308X(03)54008-8
- 300 Drummond, A., Pybus, O. G., Rambaut, A., Forrest, S. A., & Rodrigo, A. G. (2003, September
301 1). Measurably evolving populations. *Trends in Ecology and Evolution*, Vol. 18, pp. 481–
302 488. doi: 10.1016/S0169-5347(03)00216-7
- 303 Duchêne, S., Duchêne, D., Holmes, E. C., & Ho, S. Y. W. (2015). The Performance of the Date-
304 Randomization Test in Phylogenetic Analyses of Time-Structured Virus Data. *Molecular*
305 *Biology and Evolution*, 32(7), 1895–1906. doi: 10.1093/molbev/msv056
- 306 Duchene, S., Lemey, P., Stadler, T., Ho, S. Y., Duchêne, D., Dhanasekaran, V., & Baele, G.
307 (2019). Bayesian Evaluation of Temporal Signal in Measurably Evolving Populations.
308 *BioRxiv*. doi: <https://doi.org/10.1101/810697>
- 309 Duffy, S., & Holmes, E. C. (2009). Validation of high rates of nucleotide substitution in
310 geminiviruses: Phylogenetic evidence from East African cassava mosaic viruses. *Journal*
311 *of General Virology*, 90(6), 1539–1547. doi: 10.1099/vir.0.009266-0
- 312 Eldholm, V., Monteserin, J., Rieux, A., Lopez, B., Sobkowiak, B., Ritacco, V., & Balloux, F.
313 (2015). Four decades of transmission of a multidrug-resistant Mycobacterium
314 tuberculosis outbreak strain. *Nature Communications*, 6. doi: 10.1038/ncomms8119
- 315 Faria, N. R., Rambaut, A., Suchard, M. A., Baele, G., Bedford, T., Ward, M. J., ... Lemey, P.
316 (2014). The early spread and epidemic ignition of HIV-1 in human populations. *Science*,
317 346(6205), 56–61. doi: 10.1126/science.1256739

- 318 Fu, Q., Mittnik, A., Johnson, P. L. F., Bos, K., Lari, M., Bollongino, R., ... Krause, J. (2013). A
319 revised timescale for human evolution based on ancient mitochondrial genomes.
320 *Current Biology*, 23(7), 553–559. doi: 10.1016/j.cub.2013.02.044
- 321 Guindon, S., Dufayard, J. F., Lefort, V., Anisimova, M., Hordijk, W., & Gascuel, O. (2010). New
322 algorithms and methods to estimate maximum-likelihood phylogenies: Assessing the
323 performance of PhyML 3.0. *Systematic Biology*, 59(3), 307–321. doi:
324 10.1093/sysbio/syq010
- 325 Korber, B., Muldoon, M., Theiler, J., Gao, F., Gupta, R., Lapedes, A., ... Bhattacharya, T.
326 (2000). Timing the ancestor of the HIV-1 pandemic strains. *Science*, 288(5472), 1789–
327 1796. doi: 10.1126/science.288.5472.1789
- 328 Laenen, L., Vergote, V., Vanmechelen, B., Tersago, K., Baele, G., Lemey, P., ... Maes, P. (2019).
329 Identifying the patterns and drivers of Puumala hantavirus enzootic dynamics using
330 reservoir sampling. *Virus Evolution*, 5(1). doi: 10.1093/ve/vez009
- 331 Maes, P., Clement, J., Gavrillovskaya, I., & Van Ranst, M. (2004). Hantaviruses: Immunology,
332 treatment, and prevention. *Viral Immunology*, Vol. 17, pp. 481–497. doi:
333 10.1089/vim.2004.17.481
- 334 Minh, B. Q., Schmidt, H. A., Chernomor, O., Schrempf, D., Woodhams, M. D., von Haeseler,
335 A., & Lanfear, R. (2020). IQ-TREE 2: New Models and Efficient Methods for Phylogenetic
336 Inference in the Genomic Era. *Molecular Biology and Evolution*, 37(5), 1530–1534. doi:
337 10.1093/molbev/msaa015
- 338 Murray, G. G. R., Wang, F., Harrison, E. M., Paterson, G. K., Mather, A. E., Harris, S. R., ...
339 Welch, J. J. (2016). The effect of genetic structure on molecular dating and tests for
340 temporal signal. *Methods in Ecology and Evolution*, 7(1), 80–89. doi: 10.1111/2041-
341 210X.12466
- 342 Navascués, M., Depaulis, F., & Emerson, B. C. (2010). Combining contemporary and ancient
343 DNA in population genetic and phylogeographical studies. *Molecular Ecology Resources*,
344 10(5), 760–772. doi: 10.1111/j.1755-0998.2010.02895.x
- 345 O’Hanlon, S. J., Rieux, A., Farrer, R. A., Rosa, G. M., Waldman, B., Bataille, A., ... Fisher, M. C.
346 (2018). Recent Asian origin of chytrid fungi causing global amphibian declines. *Science*,
347 360(6389), 621–627. doi: 10.1126/science.aar1965
- 348 Paradis, E., & Schliep, K. (2019). ape 5.0: an environment for modern phylogenetics and
349 evolutionary analyses in R. *Bioinformatics*, 35(3), 526–528. doi:
350 10.1093/bioinformatics/bty633
- 351 R Core Development Team. (2020). R: a language and environment for statistical computing,
352 3.2.1. Document Freely Available on the Internet at: [Http://Www. r-Project. Org](http://www.r-project.org). doi:
353 10.1017/CBO9781107415324.004
- 354 Rambaut, A. (2000). Estimating the rate of molecular evolution: Incorporating non-
355 contemporaneous sequences into maximum likelihood phylogenies. *Bioinformatics*,

356 16(4), 395–399. doi: 10.1093/bioinformatics/16.4.395

357 Rambaut, A. (2020). Phylogenetic analysis of nCoV-2019 genomes. Retrieved from
358 <http://virological.org/t/356>

359 Rambaut, A., Lam, T. T., Carvalho, L. M., & Pybus, O. G. (2016). Exploring the temporal
360 structure of heterochronous sequences using TempEst (formerly Path-O-Gen). *Virus*
361 *Evolution*, 2(1). doi: 10.1093/VE/VEW007

362 Ramsden, C., Melo, F. L., Figueiredo, L. M., Holmes, E. C., & Zanotto, P. M. A. (2008). High
363 Rates of Molecular Evolution in Hantaviruses. *Molecular Biology and Evolution*, 25(7),
364 1488–1492. doi: 10.1093/molbev/msn093

365 Rieux, A., & Balloux, F. (2016, May 1). Inferences from tip-calibrated phylogenies: A review
366 and a practical guide. *Molecular Ecology*, Vol. 25, pp. 1911–1924. doi:
367 10.1111/mec.13586

368 Rieux, A., Eriksson, A., Li, M., Sobkowiak, B., Weinert, L. A., Warmuth, V., ... Balloux, F.
369 (2014). Improved Calibration of the Human Mitochondrial Clock Using Ancient
370 Genomes. *Molecular Biology and Evolution*, 31(10), 2780–2792. doi:
371 10.1093/molbev/msu222

372 Shankarappa, R., Margolick, J. B., Gange, S. J., Rodrigo, A. G., Upchurch, D., Farzadegan, H., ...
373 Mullins, J. I. (1999). Consistent Viral Evolutionary Changes Associated with the
374 Progression of Human Immunodeficiency Virus Type 1 Infection. *Journal of Virology*,
375 73(12), 10489–10502. doi: 10.1128/jvi.73.12.10489-10502.1999

376 Shapiro, B., Drummond, A. J., Rambaut, A., Wilson, M. C., Matheus, P. E., Sher, A. V., ...
377 Cooper, A. (2004). Rise and fall of the Beringian steppe bison. *Science*, 306(5701), 1561–
378 1565. doi: 10.1126/science.1101074

379 Shapiro, B., Ho, S. Y., Drummond, A., Suchard, M. A., Pybus, O. G., & Rambaut, A. (2011). A
380 Bayesian Phylogenetic Method to Estimate Unknown Sequence Ages. *Molecular Biology*
381 *and Evolution*, 28(2), 879–887. doi: 10.1093/molbev/msq262

382 Sicard, A., Zeilinger, A. R., Vanhove, M., Schartel, T. E., Beal, D. J., Daugherty, M. P., &
383 Almeida, R. P. P. (2018). *Xylella fastidiosa* : Insights into an Emerging Plant Pathogen.
384 *Annual Review of Phytopathology*, 56(1), 181–202. doi: 10.1146/annurev-phyto-
385 080417-045849

386 Sievert, C. (2020). *Interactive Web-Based Data Visualization with R, plotly, and shiny*.
387 Chapman and Hall/CRC.

388 Stamatakis, A. (2014). RAxML version 8: A tool for phylogenetic analysis and post-analysis of
389 large phylogenies. *Bioinformatics*, 30(9), 1312–1313. doi:
390 10.1093/bioinformatics/btu033

391 Stiller, M., Baryshnikov, G., Bocherens, H., Grandal d’Anglade, A., Hilpert, B., Munzel, S. C., ...
392 Knapp, M. (2010). Withering Away--25,000 Years of Genetic Decline Preceded Cave

393 Bear Extinction. *Molecular Biology and Evolution*, 27(5), 975–978. doi:
394 10.1093/molbev/msq083

395 Vanhove, M., Retchless, A. C., Sicard, A., Rieux, A., Coletta-Filho, H. D., De La Fuente, L., ...
396 Almeida, R. P. P. (2019). Genomic diversity and recombination among *Xylella fastidiosa*
397 subspecies. *Applied and Environmental Microbiology*, 85(13). doi: 10.1128/AEM.02972-
398 18

399 Weinert, L. A., Welch, J. J., Suchard, M. A., Lemey, P., Rambaut, A., & Fitzgerald, J. R. (2012).
400 Molecular dating of human-to-bovid host jumps by *Staphylococcus aureus* reveals an
401 association with the spread of domestication. *Biology Letters*, 8(5), 829–832. doi:
402 10.1098/rsbl.2012.0290

403 Yu, G., Smith, D. K., Zhu, H., Guan, Y., & Lam, T. T. (2017). ggtree: an r package for
404 visualization and annotation of phylogenetic trees with their covariates and other
405 associated data. *Methods in Ecology and Evolution*, 8(1), 28–36. doi: 10.1111/2041-
406 210X.12628

407

408

409

410

411

412

413

414

415

416

417

418

419 **Data accessibility**

420 PhyloSTemS can be executed online at <https://pvbmt-apps.cirad.fr/apps/phylostems/> but
421 source code can also be downloaded <https://gitlab.com/cirad-apps/PhyloSTemS> for local
422 implementation. The two empirical trees used in this paper (Hantaviruses and *Xylella*
423 *fastidiosa*) are accessible from the gitlab repository.

424 **Authors contribution**

425 A.R initially conceptualized the method. A.P generated a first version of the code. A.D
426 improved it and converted it into a Shiny application with advices from A.R, G.C & F.C. G.C
427 managed the online implementation & maintenance of the app. A.D & A.R wrote the first
428 draft and all authors contributed to the final version.

429

430

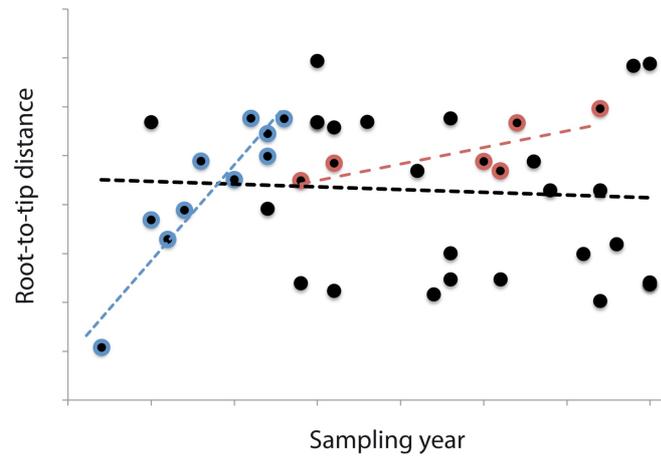
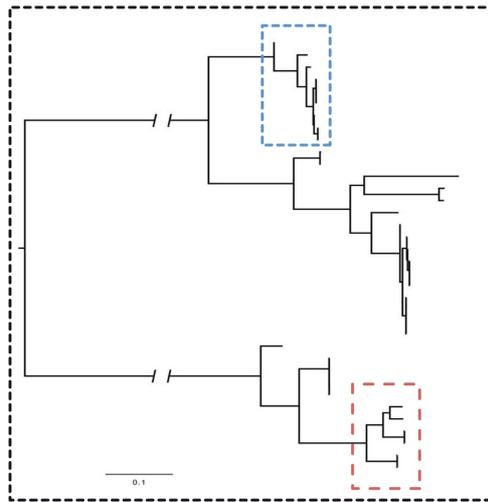
431

432

433

434

435



436

437

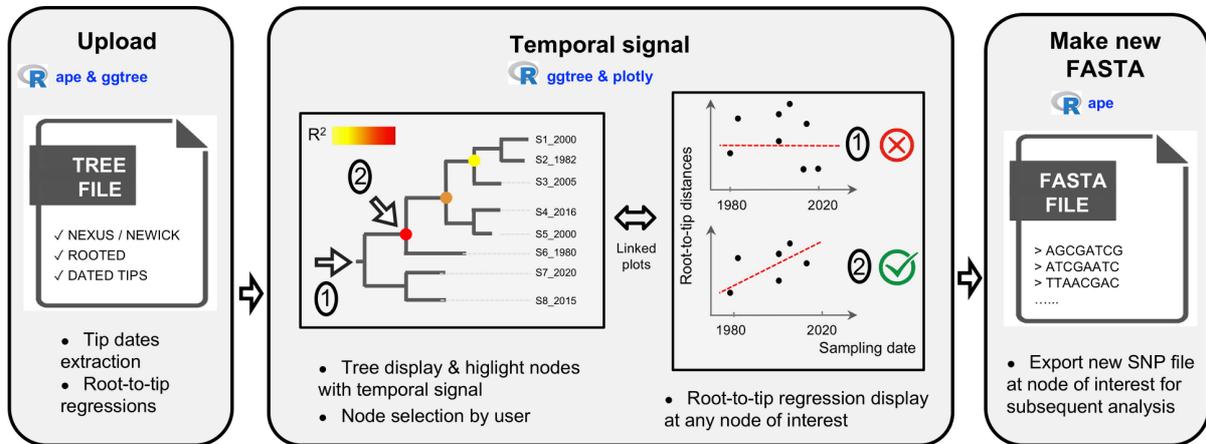
438 **Fig. 1.** Let's a tree (left panel) be constructed from a random dataset of heterochronous
 439 sequences. When investigating temporal signal on the whole dataset using the regular root-
 440 to-tip regression method (right panel), no significant signal was found as the slope of the
 441 regression (black dotted line) appears to be non-positive. Hence, tip-based inferences should
 442 not be performed at the whole data set timescale. However, as illustrated by the red and
 443 blue positive (and significant) regression slopes calculated on two subsets of samples (red
 444 and blue squares on the tree), positive temporal signal exists at reduced evolutionary
 445 timescales at which thorough tip-based inferences could be performed. The main objective
 446 of PhyloSTemS is to provide the user with a graphical tool to detect without *a priori* such
 447 evolutionary clades.

448

449

450

451



452
453 **Fig. 2.** Schematic representation of PhyloSTemS workflow. Main boxes (“Upload”, “Temporal
454 signal” & “Make new FASTA”) represent the internal structure of the application organized in
455 three main panels. Major tasks performed in each panel are summarized along with sourced
456 R packages.

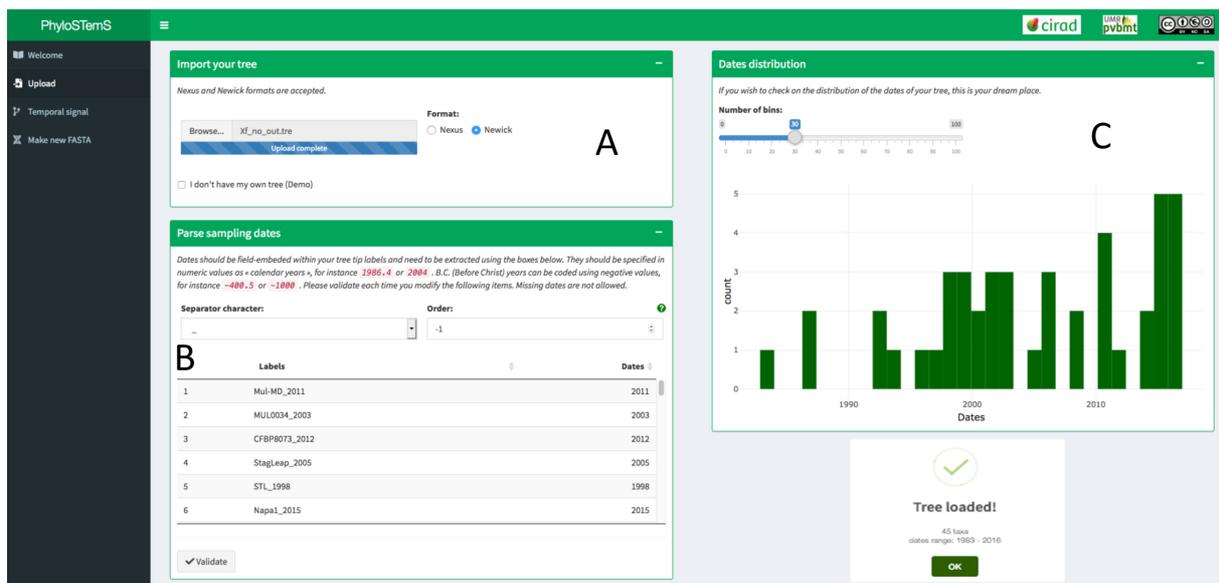
457

458

459

460

461



462

463

464 **Fig. 3.** PhyloSTemS's upload panel requesting the user to load a phylogenetic tree (A) and
 465 specify tip sampling dates from field-embedded values (B). Once loaded, a distribution of
 466 sampling dates is plotted allowing for a visual check of sequences temporal width (C).

467

468

469

470

471

472

473

474

475

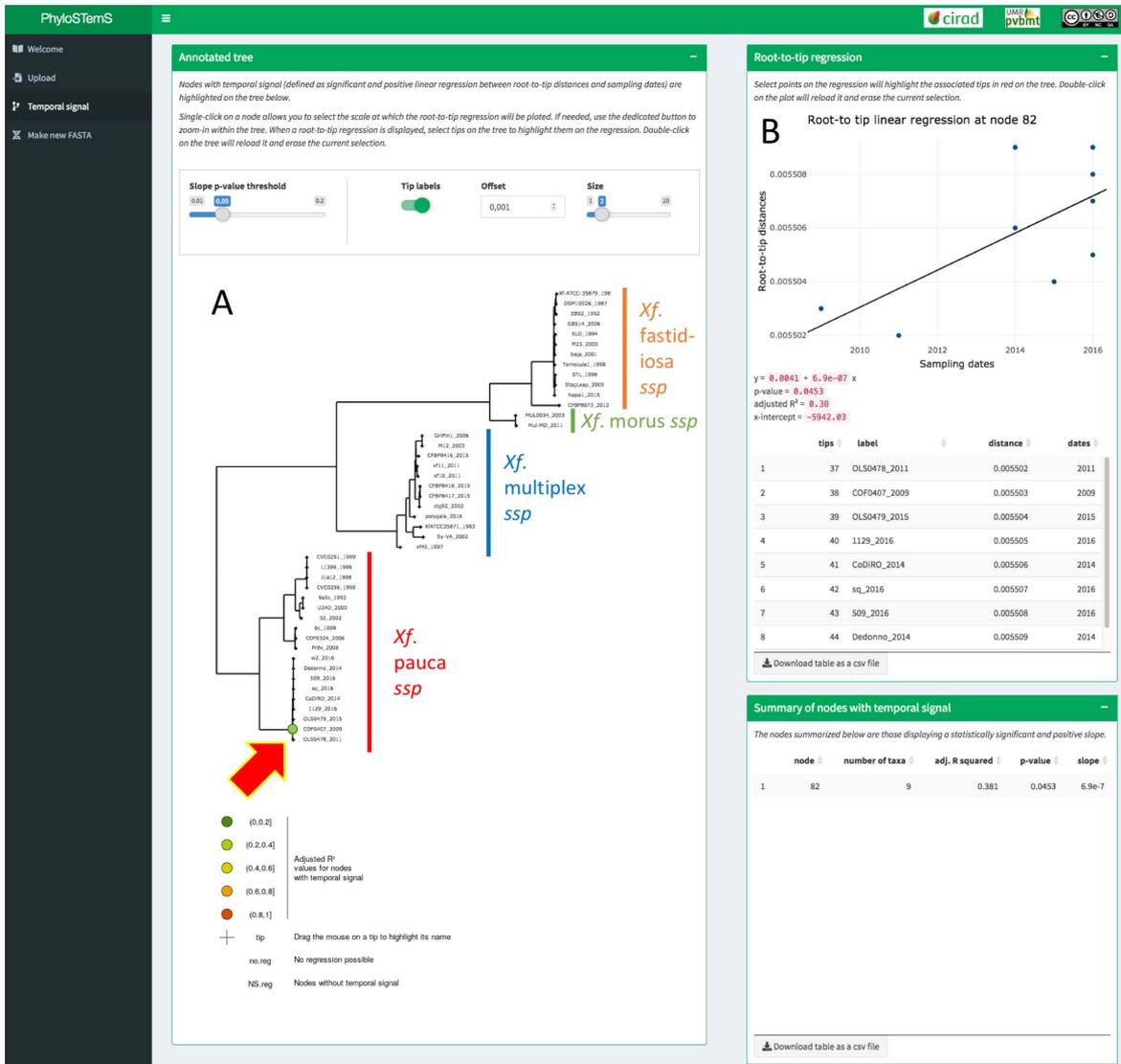
476

477

478

479

480



481

482 **Fig. 4.** Annotated phylogenetic tree of *Xylella fastidiosa* empirical dataset (A). Red arrow
 483 indicates the only node at which temporal signal was found. Root-to-tip regression at this
 484 node, along with associated parameters are plotted in (B).

485

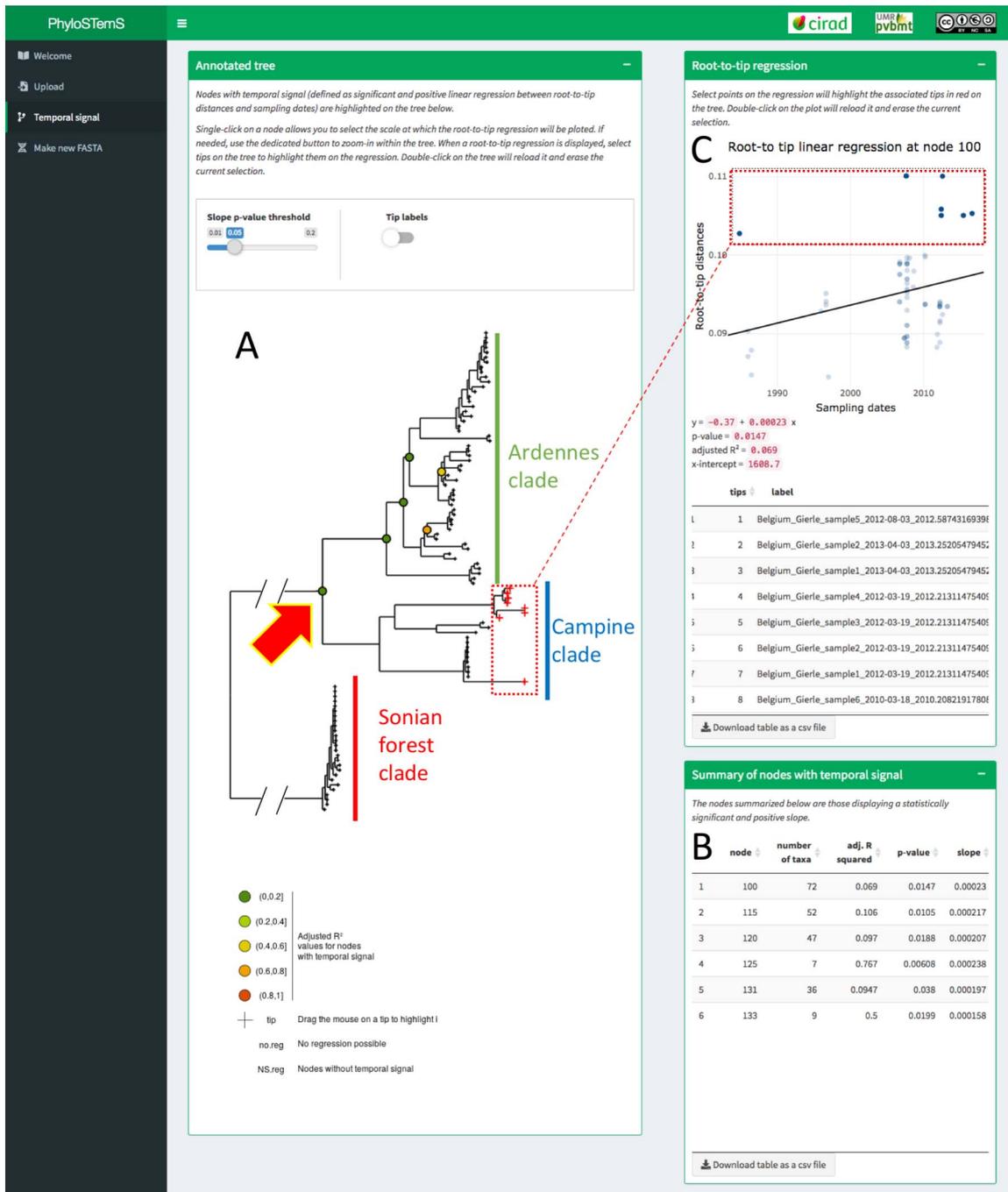
486

487

488

489

490



491

492

493 **Fig. 5.** Annotated phylogenetic tree for the Hantaviruses empirical dataset (A). Coloured
 494 circles indicate nodes at which temporal signal was found. A table summarizing those nodes,
 495 along with associated linear regression parameters is given in (B). Root-to-tip regression at
 496 node highlighted by the red arrow is plotted in (C). Both the tree and the regression plots are
 497 linked, so that data points (or tree tips) selected in one plot will automatically be highlighted
 498 on the other one, as illustrated by the red-dotted frames.

499