- DateLife: leveraging databases and analytical tools to reveal the dated Tree of Life
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DateLife: leveraging databases and analytical tools to reveal the dated Tree of Life

Abstract

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Achieving a high-quality reconstruction of a phylogenetic tree Chronograms 19 -phylogenies with branch lengths proportional to absolute time (chronogram) is a difficult and time-consuming task. But the increased availability of fossil and molecular data, and 21 time-efficient analytical techniques has resulted in many recent publications of large chronograms for a large number and wide diversity of organisms. Knowledge of the 23 evolutionary time frame of organisms is key for research in the natural sciences. It also 24 represent valuable information time—represent key data on timing of evolutionary events 25 for the study of natural processes in many areas of biological research. Chronograms also 26 provide valuable information that can be used for education, science communication, and 27 conservation policy decisions. When chronograms are shared in public and open databases, 28 this wealth of expertly-curated and peer-reviewed data on evolutionary timeframe is 29 exposed in a programatic and reusable way, as intensive and localized efforts have improved 30 data sharing practices, as well as incentivizited open science in biology Yet, achieving a 31 high-quality reconstruction of a chronogram is a difficult and resource-consuming task. 32 Here we present DateLife, a service phylogenetic software implemented as an R package and an R Shiny website web application available at www.datelife.org, that provides functionalities services for efficient and easy finding discovery, summary, reuse, and 35 reanalysis of node age data mined from a curated database of expert, peer-reviewed, public data on time frame of evolution and openly available chronograms. The main DateLife workflow constructs a chronogram for any given combination of taxon names by searching a starts with one or more scientific taxon names provided by a user. Names are processed and standardized to a unified taxonomy, allowing DateLife to run a name match across its local chronogram database constructed and curated from the that is curated from Open Tree of LifePhylesystem phylogenetic database, which incorporates phylogenetic data from

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the TreeBASE database as well. We implement and test methods for summarizing time
   data from multiple source chronograms using supertree and congruification algorithms, and
   using age data extracted from source chronograms as secondary calibration points to add
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   branch lengthsproportional to absolute time to a tree topology. DateLife will be useful's
   phylogenetic repository, and extract all chronograms that contain at least two queried
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   taxon names, along with their metadata. Finally, node ages from matching chronograms
   are mapped using the congruification algorithm to corresponding nodes on a tree topology.
   either extracted from Open Tree of Life's synthetic phylogeny or one provided by the user.
   Congruified node ages are used as secondary calibrations to date the chosen topology, with
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   or without initial branch lengths, using different phylogenetic dating methods such as
   BLADJ, treePL, PATHd8 and MrBayes. We performed a cross-validation test to compare
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   node ages resulting from a DateLife analysis (i.e., phylogenetic dating using secondary
   calibrations) to those from the original chronograms (i.e., obtained with primary
   calibrations), and found that DateLife's node age estimates are consistent with the age
   estimates from the original chronograms, with the largest variation in ages occurring
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   around topologically deeper nodes. Because the results from any software for scientific
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   analysis can only be as good as the data used as input, we highlight the importance of
   considering the results of a DateLife analysis in the context of the input chronograms.
   DateLife can help to increase awareness of the existing variation in alternative hypothesis of
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   evolutionary time disparities among alternative hypotheses of dates for the same organisms,
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   and can foster diversification events, and to support exploration of the effect of alternative
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   evolutionary timing hypotheses on the results of chronogram hypotheses on downstream
   analyses, providing a framework for a more informed interpretation of evolutionary results.
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         Keywords: Tree; Phylogeny; Scaling; Dating; Ages; Divergence times; Open Science;
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   Congruification; Supertree; Calibrations; Secondary calibrations.
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68 Word count: 5393 6797

Chronograms –phylogenies with branch lengths proportional to time– provide key data on evolutionary time frame for the study of natural processes in many areas of biological research, such as comparative analysis (Freckleton, Harvey, & Pagel, 2002; Harvey, Pagel, & others, 1991), developmental biology (Delsuc et al., 2018; Laubichler & Maienschein, 2009), conservation biology and ecology (Felsenstein, 1985; Webb, 2000), historical biogeography (Posadas, Crisci, & Katinas, 2006), and species diversification (Magallon & Sanderson, 2001; Morlon, 2014).

Building a chronogram is not an easy task. It requires obtaining and curating data a homology hypothesis to construct a phylogeny, selecting and placing appropriate calibrations on the phylogeny using independent age data points from the fossil record or other dated events, and inferring the a full dated tree; it also generally requires. All of this entails specialized biological training, taxonomic domain knowledge, and a non-negligible significant amount of research time, computational resources and funding.

Here we present the DateLife project which has the main goal of eapturing extracting
and exposing age data from published chronograms, and making these making age data
readily accessible to the a wider community for reuse and reanalysis, for in research,
teaching, and science communication and conservation policy. DateLife's core software
application is available as an R package (Sanchez-Reyes et al., 2022), and as an online
Rshiny interactive website at www.datelife.org. It features key elements for scientific
reproducibility, such as a curated, versioned, open and fully public source chronogram
database (McTavish et al., 2015), data stored and available in a computer readable that
stores data in a computer-readable format (Vos et al., 2012), automated and programmatic
ways of accessing the data and downloading the data, also in a computer-readable format
(Stoltzfus et al., 2013); and methods to summarize and compare the data.

DateLife's core software application consists of applications are implemented in the R 94 package datelife. Its current stable version v0.6.8, is available from The Comprehensive 95 R Archive Network (CRAN) repository (Sanchez-Reyes et al., 2022), and relies on 96 functionalities from various other biological R packages: ape (Paradis, Claude, & Strimmer, 97 2004), bold (Chamberlainet al., 2019, 2018), geiger (Pennell et al., 2014), msa (Bodenhofer, Bonatesta, Horejš-Kainrath, & Hochreiter, 2015), paleotree (Bapst, 2012), phyloch (Heibl, 2008), phylocomr (Ooms & Chamberlain, 2018), phytools (Revell, 2012), rotl (Michonneau, 100 Brown, & Winter, 2016), and taxize (Chamberlain, 2018; Chamberlain & Szöcs, 2013; 101 Chamberlain et al., 20192013). Figure ?? 1 provides a graphical summary of the three main 102 steps of the DateLife workflow: creating a search query, searching a database, and 103 summarizing results from the search. 104

### Creating a Search Query

DateLife starts by processing an input consisting of at least two taxonnames, which
the scientific name of at least one taxon. Multiple input names can be provided as a comma
separated character string or as tip labels on a tree. If the input is a tree, it can be provided
as a classic newick character string (Archie et al., 1986), or as a "phylo" "phylo" R object
(Paradis et al., 2004). The input tree is not required to have branch lengths, and its topology
is used in the summary steps described in the next section.

DateLife accepts scientific names that can belong to any inclusive taxonomic group (e. g., genus, family, tribe, etc.) or a binomial species name. Subspecies and variants are ignored. If an input taxon name belongs to an inclusive taxonomic group, DateLife has two alternative behaviors defined by processes input scientific names using a Taxonomic Name Resolution Service (TNRS), which increases the probability of correctly finding the queried taxon names in the chronogram database. TNRS detects, corrects and standardizes name misspellings and typos, variant spellings and authorities, and nomenclatural synonyms to a single taxonomic standard (Boyle et al., 2013). TNRS also allows to correctly choose

between homonyms, by considering other taxa provided as input to infer the taxonomic context of the "get species from taxon" flag. If the flag is active, DateLife retrieves all 121 species names within the inclusive taxonomic group following a standard taxonomy of 122 choice, and adds them to the input string. Taxonomies currently supported by DateLife 123 are-homonym. DateLife implements TNRS using the Open Tree of Life (OpenTree) unified 124 Taxonomy (OTT, Open Tree Of Life et al., 2016; Rees & Cranston, 2017), as standard, 125 storing taxonomic identification numbers (OTT ids) for further processing and analysis. 126 Other taxonomies currently supported by DateLife are the National Center of 127 Biotechnology Information (NCBI) taxonomic database (Schoch et al., 2020), the Global 128 Biodiversity Information Facility (GBIF) taxonomic backbone (GBIF Secretariat, 2022), and 129 the Interim Register of Marine and Nonmarine Non-marine Genera (IRMNG) database 130 (Rees, Vandepitte, Decock, & Vanhoorne, 2017). If et al., 2017). 131

Besides binomial species names, DateLife accepts scientific names from any inclusive taxonomic group (e.g., genus, family, tribe), as well as subspecific taxonomic variants (e.g., subspecies, variants, strains). If a taxon name belongs to an inclusive taxonomic group,

DateLife has two alternative behaviors defined by the "get species from taxon" flag. If the flag is active, DateLife retrieves all species names within a taxonomic group provided, from a standard taxonomy of choice, and adds them to the flag is inactive, DateLife excludes

any taxon names above the species level from the search query.

DateLife processes input scientific names using a Taxonomic Name Resolution Service (TNRS), which increases the probability of correctly finding the queried taxon names in the chronogram database. TNRS detects, corrects and standardizes name misspellings and typos, variant spellings and authorities, and nomenclatural synonyms to a single taxonomic standard (Boyle et al., 2013). DateLife implements TNRS with OTT as standard (Open Tree Of Life et al., 2016; Rees & Cranston, 2017), storing taxonomic identification numbers for further processing.

The processed input In this case, subspecific variants are excluded. If the flag is
inactive, DateLife excludes inclusive taxon names from the search query, and species and
subspecific variant names are processed as provided by the user. The processed taxon
names are saved as an R object of a newly defined class, datelifeQuery, that is used in the
following steps. This object contains the standardized names input names standardized to a
taxonomy of choice (OTT by default), the corresponding OTT identification id numbers,
and the topology of the input tree and input tree, if one was provided.

### Searching a Chronogram Database

At the time of writing of this manuscript (Jun 22, 2022Oct 27, 2023), DateLife's chronogram database latest version consist of 253 chronograms published in 187 different studies. It is curated from OpenTree's phylogenetic database, the Phylesystem, which constitutes an open source of expert and peer-reviewed phylogenetic knowledge with rich metadata (McTavish et al., 2015), which allows automatic and reproducible assembly of our chronogram database. Datelife's chronogram database is navigable as an R data object within the datelife R package.

A unique feature of the Phylesystem is that any user can add new published, 161 state-of-the-art chronograms any time, through their OpenTree's curator application 162 (https://tree.opentreeoflife.org/curator). As chronograms are added to Phylesystem, they are 163 can be incorporated into the chronogram database of the datelife package, 164 which is currently manually updated as new chronogram data is added to Phylesystem. 165 The updated database is assigned a new version number, followed by a package release on CRAN. Users can directly run datelife 's chronogram database is updated as new ehronogram data is added to Phylesystem, at a minimum of once a month and a maximum 168 of every 6 months. Users can also implement functions from the datelife R package 169 functions to trigger an update of the their local chronogram database, to incorporate any 170 new chronograms to the user's their DateLife analysis before an official database update is 171

released on CRAN.

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A DateLife search is implemented by matching processed taxon names provided by the 173 user to tip labels in the chronogram database. Chronograms with at least two matching 174 taxon names on their tip labels are identified and pruned down to preserve only the matched 175 taxa. These matching pruned chronograms are referred to as source chronograms. Total 176 distance (in units of millions of years million years (Myr) between taxon pairs within each 177 source chronogram are stored as a patristic distance matrix (Figure ??Fig. 1). The matrix 178 format speeds up extraction of pairwise taxon ages of any queried taxa, as opposed to 179 searching the ancestor node of a pair of taxa in a "phylo" object or newick string. 180 Finally, the patristic matrices are associated to the study citation where the original 181 chronogram was published, and stored as an R object of the newly defined class 182 datelifeResult. 183

# Summarizing Search Results

Summary information is extracted from the datelifeResult object to inform decisions for subsequent steps in the analysis workflow. Basic summary information available to the user is:

- 1. The matching pruned chronograms as newick strings or "phylo" "phylo" objects.
- 2. The ages of the root of all source chronograms. These ages can correspond to the age of the most recent common ancestor (mrca) of the user's group of interest if the source chronograms have all taxa belonging to the group. If not, the root corresponds to the mrca of a subgroup withing within the group of interest.
- 3. Study citations where original chronograms were published.
- 4. A report of input taxon names matches across source chronograms.
- 5. The source chronogram(s) with the most input taxon names.
  - 6. Various single summary chronograms resulting from summarizing age data, generated

using the methodology described next.

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# Choosing a Topology

DateLife requires a tree topology to summarize age data upon. We recommend that 199 users provide as input a tree topology from the literature, or one of their own making. If no 200 topology is provided, DateLife automatically extracts one from the OpenTree synthetic tree, 201 a phylogeny currently encompassing 2.3 million taxa across all life, assembled from 1,239 202 published phylogenetic trees and OpenTree's unified Taxonomy, OTT (Open Tree Of Life et 203 al., 2019). Alternatively, DateLife can combine topologies from source chronograms using a 204 supertree approach. To combine topologies from source chronograms into a single 205 summary (or supertree)topology, the DateLife workflow (Criscuolo, Berry, Douzery, & 206 Gascuel, 2006). To do this, DateLife first identifies the source chronograms that form a 207 grove, roughly, a sufficiently overlapping set of taxa between trees, by implementing 208 definition 2.8 for n-overlap from Ané et al. (2009). If the source chronograms do not form a 200 grove, the supertree reconstruction will fail. In rare cases, a group of trees can have multiple 210 groves. By default, DateLife chooses the grove with the most taxa, however, the "criterion = 211 trees" "criterion = trees" flag allows the user to choose the grove with the most trees instead. 212 If source chronograms do not form a grove, the supertree reconstruction will fail. The 213 result is a single summary (or supertree) topology, that combines topologies from source chronograms in a grove.

### Dating the Topology Applying Secondary Calibrations

Input topologies from OpenTree or the supertreeapproach described above do not include branch length estimates of any kind. Optionally, to estimate branch lengths proportional to substitution rates on these topologies, DateLife can mine the Barcode of Life Data System, BOLD (Ratnasingham & Hebert, 2007) to obtain genetic markers for the input taxa. These markers are aligned with MUSCLE (Edgar, 2004) (by default) or

MAFFT (Katoh, Asimenos, & Toh, 2009). This alignment can be used to estimate branch lengths on input topologies that lack branch lengths. Currently, branch length reconstruction in DateLife is performed using parsimony and the likelihood of the phylogenetic tree given a sequence alignment is computed (Schliep, 2011). While relative branch length information provides additional data for nodes without secondary date calibrations, topologies without branch lengths can also be dated.

Once a topology is chosen, DateLife applies the congruification method (Eastman, 228 Harmon, & Tank, 2013) to that find nodes belonging to the same clade across source 229 chronograms, and extract then extracts the corresponding node ages from the patristic distance matrices stored as a datelifeResult. By definition, the object. Note that by definition, these matrices store total distance (time from tip to tip), hence assuming that the 232 terminal taxa are coeval and occur at the present. Hence, node ages correspond to half the 233 values stored in the patristic distance matrices. This assumes that the terminal taxa are 234 coeval and occur at the present. datelifeResult matrices. A table of congruified node 235 ages that can be used as calibrations for a dating analysis is stored as a 236 congruifiedCalibrations object. 237

For each congruent node, the pairwise distances that traverse that node are 238 summarized into a single summary matrix using classic summary statistics (i.e., mean, 239 median, minimum and maximum ages), and the Supermatrix Distance Method (SDM; 240 Criscuolo, Berry, Douzery, and Gascuel (et al., 2006), which deforms patristic distance 241 matrices by minimizing variance and then averaging them. These single summary taxon pair age matrices (Summarized calibrations) can be applied as are stored as summarized 243 calibrations that can be used as secondary calibrations to date a tree topology, using different - with or without initial branch lengths, using phylogenetic dating methods 245 currently supported within DateLife: BLADJ (Webb, Ackerly, & Kembel, 2008; Webb & 246 Donoghue, 2005), MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), PATHd8 (Britton, Anderson, Jacquet, Lundqvist, & Bremer, 2007), BLADJ (Webb, Ackerly, Webb, Ackerly, Webb, Webb & Donoghue, 2005), and treePL (Smith & O'Meara, 2012).

By default

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# Dating a Tree Topology

Dating a tree without branch lengths. - To date a tree topology when branch 252 lengths are unavailable, DateLife implements the Branch Length Adjuster (BLADJ) 253 algorithm to obtain a fully dated topology. BLADJ fixes node ages that have calibration 254 data, and distributes time between (Webb et al., 2008; Webb & Donoghue, 2005), which 255 only requires a tree topology with no branch lengths and at least two node ages to use as 256 calibrations, one for the tree root and one for any internal node of the topology. The 257 BLADJ algorithm fixes ages for nodes with calibration data upon the given tree topology. 258 Then, it assigns ages to nodes with no data evenly between nodes with calibration data. 259 This minimizes available age information by distributing time evenly between calibrated 260 nodes, minimizing age variance in the resulting chronogram. This approach has proven 261 useful for ecological analyses that require a phylogenetic time context (Webb et al., 2008). 262 BLADJ does not use branch lengths even when they are present in the input tree or 263 summarizing topology. When there is conflict in ages between nodes with calibration 264 data between ages of calibrated nodes, BLADJ ignores node ages that are older than the age 265 of a parent node. BLADJ The BLADJ algorithm requires a root age estimate. If to run. 266 Users can provide an appropriate root age estimate of their own or one obtained from the 267 literature. If a root age is not provided and there is no information on the age of the root in the chronogram database, users can provide an estimate from the literature. If none is 269 provided, DateLife assigns an arbitrary age to the root as 10% older than the oldest age available within the group DateLife chooses a random age for the root, so that a dated tree 271 topology can be generated with BLADJ. In this case, DateLife will provide a conspicuous 272 warning message, so that users are aware that the root of the chronogram was chosen at 273

random because there was no information available for it in the chronogram database,
along with suggestions on how the user can find and provide an appropriate age for the
root of the initial topology.

An alternative to BLADJ to date tree topologies in the absence of initial branch 277 lengths that is common practice in the literature is to use a birth-death model to draw 278 branch lengths (Jetz, Thomas, Joy, Hartmann, & Mooers, 2012; Rabosky et al., 2018; 279 Smith & Brown, 2018). In addition to the initial tree topology and nodes with age data, 280 these methods require initial values of speciation and extinction rate parameters provided 281 by the user. DateLife implements this approach with MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), using nodes with published age data as calibration priors on nodes of a tree topology with no branch lengths, a simple birth-death model with 284 speciation and extinction rate parameters that are provided by the user, and no genetic 285 data. However, BLADJ is the default option in DateLife, as it does not require any 286 information on diversification rates for the phylogenetic sample to draw from a branch 287 length distribution. 288

**Dating a tree with branch lengths.** – Relative branch lengths can provide key 289 information for phylogenetic dating, specifically for nodes without any calibration data 290 available. While using initial branch length data is the golden standard for phylogenetic 291 dating analyses, estimating trees with branch lengths proportional to substitution rates per 292 site requires obtaining primary data, assembling and curating a homology (orthology) 293 hypothesis, and choosing and implementing a method for phylogenetic inference. DateLife 294 implements a workflow to streamline this process by applying open data from the Barcode 295 of Life Data System, BOLD (Ratnasingham & Hebert, 2007) to obtain genetic markers for 296 input taxa. By default, BOLD genetic sequences are aligned with MUSCLE (Edgar, 2004) 297 using functions from the msa R package (Bodenhofer et al., 2015). Alternatively, sequences 298 can be aligned with MAFFT (Katoh, Asimenos, & Toh, 2009), using functions from the 299

ape R package (Paradis et al., 2004). The BOLD sequence alignment is then used to 300 obtain initial branch lengths with the accelerated transformation (ACCTRAN) parsimony 301 algorithm, which resolves ambiguous character optimization by assigning changes along 302 branches of the tree as close to the root as possible (Agnarsson & Miller, 2008), resulting in 303 older internal nodes as compared to other parsimony algorithms (Forest et al., 2005). The 304 parsimony branch lengths are then optimized using Maximum Likelihood, given the 305 alignment, the topology and a simple Jukes-Cantor model, producing a BOLD tree with 306 branch lengths proportional to expected number of substitutions per site. Both parsimony 307 and ML optimizations are done with functions from the phangorn package (Schliep, 2011). 308 Due to the computing load it requires, the BOLD workflow is currently only supported 309 through DateLife's R package. It is not yet available through the web application. 310

Alternative phylogenetic dating options Phylogenetic dating methods supported in 311 DateLife (MrBayes, PATHD8, TreePL) that incorporate branch length information from the 312 input topology in combination with the calibrations. secondary calibrations include: 313 PATHd8is, a non-clock, rate-smoothing method to date trees (Britton et al., 2007) to date 314 trees.; treePL (Smith & O'Meara, 2012), is a semi-parametric, rate-smoothing, penalized 315 likelihood dating method (Sanderson, 2002). The; and MrBayes (Huelsenbeck & Ronquist, 316 2001; Ronquist & Huelsenbeck, 2003) approach in DateLife uses the calibrations as priors on node ages, a Bayesian inference program implementing Markov chain Monte Carlo 318 (MCMC) methods to estimate a posterior distribution of model parameters. 319

#### Visualizing Results

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Finally, users can save all source and summary chronograms in formats that permit
reuse and reanalyses (newick and R "phylo" format), as well as visualize and compare
results graphically, or allowing for reuse and reanalysis, such as newick and the R "phylo"
format. Input and summary chronograms can be visualized and compared graphically, and
users can construct their own graphs using DateLife's chronogram plot generation functions

available from the R package datelifeplot (Sanchez-Reyes & O'Meara, 2022).

327 BENCHMARK

R package datelife 's R package code speed was tested on an Apple iMac with one 328 3.4 GHz Intel Core i5 processor. We registered variation in computing time of query 329 processing and search through the database relative to number of queried taxon names. 330 Query processing time increases roughly linearly with number of input taxon names, and increases considerably if Taxonomic Name Resolution Service (TNRS) is activated. Up to 332 ten thousand names can be processed and searched in less than 30 minutes with the most 333 time consuming settings. Once names have been processed as described in methods, a name 334 search through the chronogram database can be performed in less than a minute, even with 335 a very large number of taxon names (Fig. ???2). 336

datelife's code performance was evaluated with a set of unit tests designed and implemented with the R package testthat (R Core Team, 2018) that were run both locally with the devtools package (R Core Team, 2018), and on a public server using the continuous integration tool of GitHub actions (https://docs.github.com/en/actions). At present, unit tests cover more than 40% of datelife's code (https://codecov.io/gh/phylotastic/datelife). Unit testing helps identify potential issues as code is updated or, more critically, as services code relies upon may change.

Case Studies

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We illustrate the DateLife workflow using a family within the Passeriform passeriform
birds encompassing the true finches, Fringillidae, as case study. On a small example, we
analysed 6 bird species, and results from each step of the workflow are shown in Fig. ??.

Figure 3. As a second example, we analysed 289 bird species in the family Fringillidae that
are included in the NCBI taxonomy. The resulting summary chronogram summary
chronogram resulting from the DateLife analysis is shown in Fig. ??Figure 5, and results

from previous steps of the workflow are available as Supplementary Figures.

## A Small Example

Creating a search query.—We chose 6 bird species within the Passeriformes. The 353 sample includes two species of cardinals: the black-thighed grosbeak – Pheucticus tibialis 354 and the crimson-collared grosbeak – Rhodothraupis celaeno; three species of buntings: the 355 yellowhammer – Emberiza citrinella, the pine bunting – Emberiza leucocephalos and the 356 yellow-throated bunting – Emberiza elegans; and one species of tanager, the vegetarian finch – 357 Platyspiza crassirostris. Processing of input names found that Emberiza elegans is synonym 358 for Schoeniclus elegans in the default reference taxonomy (OTT v3.3, June 1, 2021). For a 359 detailed discussion on the state of the synonym, refer to Avibase (Avibase, 2022; Lepage, 360 2004; Lepage, Vaidya, & Guralnick, 2014). Discovering this synonym allowed assigning five 361 age data points for the parent node of *Emberiza elegans*, shown as *Schoeniclus elegans* in figure ?? A Figure 3a, which would not have had any data otherwise.

Searching the database.— DateLife used the processed input names to search the local chronogram database and found 9 matching chronograms in from 6 different studies (Fig. ??B3b). Three studies matched five input names (Barker, Burns, Klicka, Lanyon, & Lovette, 2015; Hedges, Marin, Suleski, Paymer, & Kumar, 2015; Jetz, Thomas, Joy, Hartmann, & Mooers, et al., 2012), one study matched four input names (Hooper & Price, 2017) and two studies matched two input names (Barker, Burns, Klicka, Lanyon, & Lovette, 2013; Burns et al., 2014). No studies matched all input names. Together, source chronograms provide 28 unique age data points, covering all nodes on our chosen tree topology to date (Table ??1).

Summarizing search results.— DateLife obtained OpenTree's synthetic tree topology for these taxa (Fig. ??C3c), and congruified and mapped age data to nodes in this chosen topology(Table ??).—, shown in Table 1. The name processing step allowed including

five data points for node "n4" "n4" (parent of Schoeniclus elegans; Fig. ??A3A) that would 376 not have had any data otherwise due to name mismatch. Age summary statistics per node 377 were calculated (Table ??2) and used as calibrations to date the tree topology using the 378 BLADJ algorithm. As expected, more inclusive nodes (e.g., node "n1" "n1") have more 379 variance in age data than less inclusive nodes (e.g., node "n5"). Summary "n5"). Median 380 summary age data for node "n2" were "n2" was excluded as final calibration because they 381 are older than age data of the it is older than the median age of a more inclusive node, "n1" 382 "n1" (Fig. ??C43c4). 383

# An Example with the Family of True Finches

Creating a query. To obtain ages for all species within the family of true finches, 385 Fringillidae, we ran a DateLife query using the "get species from taxon" get species from 386 taxon" flag, which gets all recognized species names within a named group from a taxonomy 387 of choice. Following the NCBI taxonomy, our DateLife query has 289 Fringillidae species 388 names. This taxon-constrained approach implies that the final results of a full DateLife 389 analysis will be done performed using a tree topology and ages for the species in a named 390 available for species names from a given taxonomic group, which do not necessarily 391 correspond to a monophyletic group. Users can change this behaviour by providing a 392 monophyletic tree-behavior by providing all species names corresponding to a monophyletic 393 group as input for a DateLife search, or as a tree topology for a monophyletic tree to 394 construct a DateLife summary. 395

Searching the database. Next, we used the processed species names in our 396 DateLife query to identify chronograms with at least two Fringillidae species as tip taxa. 397 The DateLife search identified 13 chronograms containing at least two Fringillidae species 19 398 chronograms matching this criteria, published in 9-13 different studies (Barker et al., 2013, 399 2015; Burns et al., 2014; Claramunt & Cracraft, 2015; Gibb et al., 2015; Hedges et al., 2015; Hooper & Price, 2017; Jetz et al., 2012; Kimball et al., 2019; Oliveros et al., 2019; Price et 401 al., 2014; Roquet, Lavergne, & Thuiller, 2014; Uyeda, Pennell, Miller, Maia, & McClain, 2017). Once identified, DateLife pruned these matching chronograms to keep Fringillidae species names on tips only remove tips that do not belong to the queried taxon names, and 404 transformed these pruned chronograms to pairwise distance matrices, revealing 1,206 405 different age data points available for species within the Fringillidae (Supplementray 406 Supplementary Table S1). 407

Summarizing search results.— The final step is to congruify and summarize

entailed congruifying and summarizing the age data available for the Fringillidae species

into two single summary chronograms, using two different types of summary ages, median and SDM. As explained in the "Description" "Description" section, a tree topology to summarize age data upon is required. By default, DateLife uses the topology from OpenTree's synthetic tree that contains the species in the search queryto summarize age data uponall taxa from the search query. According to OpenTree's synthetic tree, species belonging to the family Fringillidae do not form a monophyletic group (Fig. ??).

Age data from source chronograms was congruified to OpenTree's topology 4a).

Hence, a topology containing only the 289 species from the original query was extracted

from Open Tree of Life's synthetic tree v12.3 (Fig. ??B)4b; Open Tree Of Life et al., 2019).

Source chronograms (Supplementary Figs. S2-S20) were congruified to OpenTree's 420 topology shown in Figure 4b, reducing the original 1,206 node age data set to 818 different 421 data points (Supplementary Supplementary Table S2). For each congruent node, age 422 summary statistics were calculated and used as fixed secondary calibrations over the 423 chosen tree topology, that can be used as calibrations for the chosen topology (Fig. 4b). 424 The congruent node age data points were summarized for each node, resulting in 194 425 summary node ages. From these 21 were excluded as secondary calibrations because they were older than the ancestral node. The remaining 173 summary node ages were used as 427 secondary calibrations to obtain a fully dated (and resolved) phylogeny with the program 428 BLADJ (Fig. ??5).

#### CROSS-VALIDATION TEST

We performed a cross validation analysis of the DateLife workflow test of a DateLife
analysis using the Fringillidae chronograms. We used the source chronograms obtained
above (Supplementary Figs. S2-S20). We used as inputs for a DateLife analysis all
individual tree topologies from each of the 19 source chronograms from 13 studies as inputs,

treating their node ages as unknown. We then estimated dates for these topologies using 435 the node ages congruified node ages extracted from chronograms from the chronograms 436 from the other studies as calibrations and smoothing using BLADJ-all other studies upon 437 the individual topologies, effectively excluding original ages from each topology. Finally, 438 average node ages per node were applied as secondary calibrations and smoothed with the 439 BLADJ algorithm. We found that node ages from original study the original studies, and 440 ages estimated using all other age data available are largely correlated (Fig. ??6). For five 441 studies, Datelife DateLife tended to underestimate ages for topologically deeper nodes (those with many descendant taxa, aka 'closer to the root' (closer to the root) relative to the 443 original estimate, and overestimate ages for nodes closer to the tips. Accordingly, root ages 444 are generally older in the original study than estimated using cross-validated ages 445 (Supplementary Fig. S1). In general, topologically deeper nodes display the largest age variation between node ages from the original chronograms and ages summarized with DateLife.

DISCUSSION

DateLifemakes 's goal is to improve availability and accessibility of state-of-the-art
data on evolutionary time frame easily accessible for comparison, reuse, and reanalysis, to
researchers in of organisms, to allow users from all areas of science and with all levels of
expertise in the matter. It is to compare, reuse, and reanalyse expert age data for their
own applications. As such, it is designed as an open service that does not require any expert
biological knowledge from users—besides the scientific names of the species or group they
users want to work with, for to use any of its functionality functionalities.

A total of 99,474 unique terminal taxa are represented in DateLife's database.

Incorporation of more chronograms into the database will continue to improve DateLife's

services. One option to increase the number of chronograms in the DateLife database is the

Dryad data repository. Methods to automatically mine chronograms from Dryad could be

designed and implemented. However, Dryad's metadata system has no information to
automatically detect branch length units, and those would still need to be determined
manually by a human curator. We would like to emphasize on the importance of sharing
chronogram data, including systematically curated metadata, into open repositories, such as
OpenTree's Phylesystem (McTavish et al., 2015) for the benefit of the scientific community
as a whole.

As we envision that DateLife will have many interesting applications in research and 467 beyond, we emphasize that DateLife's results –as well as any insights gleaned from them, 468 largely depend on the quality of the source chronograms: low quality chronograms will 460 produce low quality results. The "garbage in, garbage out" problem has long been 470 recognised in supertree methods for summarizing phylogenetic trees (Bininda-Emonds et 471 al., 2004). We note that this is a surfacing issue of any automated tool for biological data 472 analysis. For example, DNA riddled with sequencing errors will produce generally poor 473 alignments that will return biased evolutionary hypothesis, independently of the quality of 474 the analysis software used. Again, we urge readers and DateLife users to explore all input 475 chronograms before using a summary chronogram resulting from a DateLife workflow. 476

Finally, uncertainty and variability of chronogram node age estimates might pose
larger issues in some research areas than others. For example, in ecological and
conservation biology studies, it has been shown that incorporating some chronogram data
provides better results than when not using any age data at all, even if the node ages are
not good quality (Webb et al., 2008). In the following sections we discuss the particularities
of divergence times from DateLife's summary chronograms and their impact on certain
evolutionary analyses, for consideration of the readers and users in different research areas.

#### Age Variation in Source Chronograms

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Conflict in estimated ages among alternative studies is common in the literature. See,

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for example, the robust ongoing debate about crown group age of angiosperms
   (Barba-Montoya, Reis, Schneider, Donoghue, & Yang, 2018; Magallón, Gómez-Acevedo,
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   Sánchez-Reves, & Hernández-Hernández, 2015; Ramshaw et al., 1972; Sanderson & Doyle,
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   2001; Sauquet, Ramírez-Barahona, & Magallón, 2021). Source-Alternative source
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   chronograms available for the same organisms taxa have potentially been estimated
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   implementing ealibrations very differently different types of calibrations, which affects the
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   resulting node age estimates. For example, the chronograms from in the DateLife analysis
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   of the Fringillidae shown above, the chronograms from one study (Burns et al. 4, 2014) were
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   inferred using molecular substitution rate estimates across birds (Weir & Schluter, 2008),
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   and have much older age estimates for the same nodes than chronograms that were inferred
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   using fossils as fossil calibrations (Figs. ??, ?? 5, 6; Supplementary Figs. S1, S5).
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Different calibrationimplementations might also S1c, S4). Another source of conflict 497 in estimated node ages can arise from different placements for the same calibration, which 498 would imply fundamentally distinct evolutionary hypotheses (Antonelli et al., 2017). For 499 example, two independent researchers working on the same clade should both carefully select 500 and justify their choices of fossil calibration placement. Yet, if one researcher concludes that 501 a fossil should calibrate the ingroup of a clade, while another researcher concludes that the 502 same fossil should calibrate the outgroup of the clade, the resulting age estimates will differ, 503 as the placement of calibrations as stem or crown group has been proven is known to 504 significantly affect estimates of time of lineage divergence estimates (Sauguet, 2013). Finally, 505 placement of calibrations also affects uncertainty of node age estimates. For example, 506 nodes that are sandwiched between a calibrated node and a calibrated root have less 507 freedom of movement and hence narrower confidence intervals (Vos & Mooers, 2004), which 508 inflates precision for nodes without calibrations but does not necessarily improve accuracy 509 of the estimated ages. 510

DateLife's summary chronograms are intended to represent all variation in estimated

node ages from source chronograms. Node age distribution ranges allow to visually explore
ages from source chronograms individually and contextualize and compare them against
other chronograms. Researchers that wish to use summary chronograms in downstream
evolutionary analysis may select multiple trees sampled from the summary distribution of
node ages, to account for variation in source chronograms.

# Primary vs Secondary Calibrations

While most chronograms in DateLife's database are constructed using primary 518 ealibrations (molecular substitution rates or ages obtained from the fossil record or 519 geological events), DateLife summarizes chronogramsusing secondary calibrations (ages 520 coming from other chronograms). Graur and Martin (2004) cautioned on the increased 521 error and uncertainty in estimated ages when using secondary calibrations in dating 522 analyses. Schenk (2016) showed that, in simulations, divergence times inferred using 523 secondary calibrations are significantly younger than those inferred with primary 524 calibrations, when obtained with Bayesian inference methods, and when priors are 525 implemented in similar ways in both analyses. Accordingly DateLife constructs summary 526 chronograms using node ages extracted from existing chronograms, i.e. secondary 527 calibrations. In general, the scientific community seems to have has more confidence in 528 chronograms obtained using primary calibrations, where the dated tree is generated from a 529 single analysis, using fossil data as primary sources of calibrations (where carefully chosen 530 fossil calibrations are the source of absolute time information, than in analyses dated using 531 secondary calibrations (Antonelli et al., 2017; Garzón-Orduña, Silva-Brandão, Willmott, 532 Freitas, & Brower, 2015; Graur & Martin, 2004; Sauguet, 2013; Sauguet et al., 2012; 533 Schenk, 2016; Shaul & Graur, 2002). However, implementation of primary calibrations is difficult: it requires specialized expertise and training to discover, place and apply 535 calibrations appropriately (Hipsley & Müller, and using 2014; Ksepka et al., 2011). One 536 approach is to use fossils that have been widely discussed and previously curated as 537

calibrations to date other trees -(Ksepka et al., 2011; Sauguet, 2013), and making sure that 538 all data reflect a coherent evolutionary history (Sauguet, 2013), as for example done by 539 Antonelli et al. (2017). There have been attempts to create fossil calibration databases. The 540 Fossil Calibration Database provides data for 220 primary calibration points encompassing 541 flowering plants and metazoans, that have been curated by experts and used for dating 542 analysis in peer-reviewed publications (Ksepka et al., 2015), though these still have room to 543 grow. This database facilitates the use of expert primary fossil calibrations in new 544 phylogenetic dating analyses. Yet, users still require the expertise to locate and calibrate 545 appropriate nodes in their phylogenies which correspond with fossils available in the 546 database. 547

It seems that using several-Recently, Powell, Waskin, and Battistuzzi (2020) showed 548 in a simulation study that secondary calibrations using node ages based on previous 549 molecular clock analyses can be as good as primary calibrations. Using several secondary 550 calibrations (as opposed to just a few) secondary calibrations one) can provide sufficient 551 information to alleviate or even neutralize potential biases (Graur & Martin, 2004; Sauguet, 552 2013). Certainly, further studies are required to fully understand the effect of secondary 553 calibrations on outputs from different tree dating methods, and on downstream analyses. It 554 is possible that secondary calibrations can be safely used with dating methods that do not 555 require setting priors, such as penalized likelihood (Sanderson, 2003), with methods that 556 do not make any assumptions on the ages and fix them to a node on a tree topology, such 557 as BLADJ (Webb et al., 2008; Webb & Donoghue, 2005), or methods that summarize age 558 data unto a tree topology. 559

; Shaul & Graur, 2002). Our cross validation analysis might provide some insight in this regard. When ages are estimated with secondary calibrations, also provides insight into the application of secondary calibrations. Node ages summarized with DateLife and those from the original studies are well correlated (Supplementary Figs. S2-S20). We also

note that DateLife estimates for nodes closer to the root do-tend to be slightly younger than 564 ages estimated with primary calibrations. Howeverfrom the original studies. In contrast, 565 nodes closer to the tip tips tend to be slightly older when estimated using secondary 566 calibrations with a dating method that does not make any prior assumptions on the nature 567 of the calibrations themselves (Supplementary Figures S2-S20)our secondary calibrations 568 than ages from the original studies. The only exception to tijs was observed on results of 560 the cross validation analysis of the exception to this trend was observed in Burns et al. 570 (2014) chronogram, which results in generally displays much younger node ages when 571 estimated using secondary calibrations than the original study (Supplementary Figs. S1, S5). 572 , supporting previous observations (Sauguet et al., 2012; Schenk, 2016). However, these 573 younger dates are more likely an example of how multiple secondary calibrations can 574 correct erroneous estimates, as dates on the Burns et al. (2014) tree were obtained using a single secondary calibration based on a previously estimated molecular evolution rate 576 across birds from Weir and Schluter (2008), and appear as major outliers compared to 577 alternate estimates for the same nodes based on primary fossil calibrations (Fig. 5). 578

Further research is needed to fully understand the effects of using secondary calibrations and the use of resulting chronograms in downstream analyses (Hipsley & Müller, 2014; Powell et al., 2020; Schenk, 2016; Shaul & Graur, 2002).

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### Sumarizing Chronograms

By default, DateLife currently summarizes all source chronograms that overlap with at least two species names. Users can exclude source chronograms if they have reasons to do so.

Strictly speaking, a good chronogram should reflect the real time of lineage divergence accurately and precisely. To our knowledge, there are no tested measures to determine independently when a chronogram is better than another. Yet, several characteristics of the data used for dating analyses, as well as from the output chronogram itself, could be used to score the quality of source chronograms.

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Some measures that have been proposed are the proportion of lineage sampling and the 590 number of calibrations used (Magallón, 2010; Magallón et al., 2015). Some characteristics 591 that are often cited in published studies as a measure of improved age estimates as compared 592 to previously published estimates are: quality of alignment (missing data, GC content), 593 lineage sampling (strategy and proportion), phylogenetic and dating inference method, 594 number of fossils used as calibrations, support for nodes and ages, and magnitude of 595 confidence intervals. 596

DateLife provides an opportunity to capture concordance and conflict among date estimates, which can also be used as a metric for chronogram reliability. Its open database of 598 chronograms allows other researchers to do such analyses themselves reproducibly, and without needing permission. Though, of course, they should follow proper citation practices, 600 especially for the source chronogram studies.

The exercise of summarizing age data from across multiple studies provides the 602 opportunity to work with a more inclusive chronogram, chronogram that reflects a unified 603 evolutionary history for a lineage, by putting together evidence from different hypotheses. 604 The largest, and taxonomically broadest chronogram currently available from OpenTree was 605 constructed summarizing age data from 2,274 published chronograms using NCBI's taxonomic tree as backbone (Hedges et al., 2015). A summarizing exercise may also amplify the effect of uncertainty and errors in source data, and blur parts of the evolutionary history 608 of a lineage that might only be reflected in source chronograms and lost on the summary 609 chronogram (Sauquet et al., 2021). 610

# Effects of Taxon Sampling on Downstream Analyses

For downstream analyses, using alternative chronogram may deeply affect our 612 inferences (Title & Rabosky, 2016), particularly when studying phenomena dependent on 613 the timing of species diversification events, such as macroevolutionary processes. 614

In ecology and conservation biology, incorporating at least some data on lineage 615 divergence times represents a relevant improvement for testing alternative hypothesis using 616 phylogenetic distance (Webb-Analysis of species diversification of simulated and empirical 617 phylogenies suggest that using a more completely sampled phylogeny provides estimates 618 that are closer to the true diversification history than when analysing incompletely 619 sampled phylogenies (Chang, Rabosky, & Alfaro, 2020; Cusimano, Stadler, & Renner, 2012; 620 Sun et al., 2008).—2020). Ideally, phylogenies should be completed using genetic data, but 621 this is a time-consuming and difficult task to achieve for many biological groups. Hence, 622 DateLife's workflow features different ways of estimating node ages in assigning divergence 623 times to taxa with missing the absence of branch length data and calibrations and branch 624 length information lengths for certain taxa. 625

Making up branch lengths Completing a phylogeny using a stochastic birth-death 626 polytomy resolver and a backbone taxonomy is a common practice in scientific publications: 627 Jetz et al. (2012), created a chronogram of all 9,993 bird species, where 67% had molecular 628 data and the rest was simulated; Rabosky et al. (2018) created a chronogram of 31,536 629 ray-finned fishes, of which only 37% had molecular data; Smith and Brown (2018) 630 constructed a chronogram of 353,185 seed plants where only 23% had molecular data. These 631 stochastically resolved chronograms return diversification rates estimates that appear less 632 biased than those estimated from their incompletely sampled counterparts, even with 633 methods that account for missing lineages by using sampling fractions (Chang et al., 2020; 634 Cusimano et al., 2012), but can also introduce spurious patterns of early bursts of 635 diversification (Cusimano & Renner, 2010; Sun et al., 2020). 636

Notably, Taxonomy-based stochastic polytomy resolvers also introduce topological
differences in phylogenetic trees. The study of macroevolutionary processes largely depends
on an understanding of the timing of species diversification events, and different
phylogenetic and chronogram hypothesis can provide very different overviews of the

macroevolutionary history of a biological group. For example, alternative topologies in 641 chronograms from the same biological group can infer very different species diversification 642 patterns (Rabosky, 2015; Title & Rabosky, 2016). Similarly, there are worries that patterns 643 of morphological evolution cannot be accurately inferred with phylogenies that have been 644 resolved stochastically over a taxonomic backbone, as any patterns would be erased by 645 randomization (Rabosky, 2015). We note that the same applies for geography- and 646 morphology-dependent diversification analysis. Hence, we suggest that phylogenies that 647 have been processed with taxonomy-based stochastic polytomy resolvers, including certain 648 summary chronograms from a DateLife analysis, can be useful as null or neutral models, 649 representing the case of a diversification process that is independent of traits and 650 geographical scenario. 651

Taxonomy-based stochastic polytomy resolvers have been used to advance research in evolution, still, risks come with this practice. Taken to the extreme, one could make generate a fully resolved, calibrated tree of all modern and extinct taxa using a single taxonomyand, a single calibration, using polytomy resolution and branch estimation methods. There has yet to be a thorough analysis of what can go wrong when one extends inferences beyond the data in this way, so we urge caution; we also urge readers and assigning branch lengths following a birth-death diversification model. Clearly, this can lead to a misrepresentation of the true evolutionary history. We urge DateLife users to follow the example of the large tree papers cited above, by carefully considering the statistical assumptions being made, potential biases, and assessing the consistency of the DateLife's results with prior work.

Conclusions

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Knowledge of the evolutionary time frame of organisms is key to many research areas: trait evolution, species diversification, biogeography, macroecology and more. It is also crucial for education, science communication and policy, but generating chronograms is difficult, especially for those who want to use phylogenies but who are not systematists, or do not have the time to acquire and develop the necessary knowledge and skills to construct
them on their own. Importantly, years of primarily public publicly funded research have
resulted in vast amounts of chronograms that are already available on in scientific
publications, but hidden to functionally hidden from the public and scientific community for
reuse.

The DateLife project allows for easy and fast summary summarization of public and 672 state-of-the-art data on time of lineage divergence. It is available as an R package, and as a 673 web-based R shiny application at www.datelife.org. DateLife provides a straightforward 674 way to get an informed idea on picture of the state of knowledge of the time frame of 675 evolution of different regions of the tree of life, and allows identification of identifying 676 regions that require more research, or that have conflicting information. It is available as an 677 R package, and as a web-based R shiny application at www.datelife.org Both Additionally, 678 both summary and newly generated trees using the DateLife workflow are useful to evaluate 679 evolutionary hypotheses in different areas of research. The DateLife project helps with We 680 hope that the DateLife project will increase awareness of the existing variation in expert estimations of time of divergence data, and will, and foster exploration of the effect of alternative divergence time hypothesis hypotheses on the results of analyses, nurturing a 683 culture of more cautious interpretation of evolutionary results.

685 AVAILABILITY

The DateLife software is free and open sourceand it. It can be used online through its
R shiny web application hosted at http://www.datelife.org, and locally through the
datelife R package, and through Phylotastic's project web portal available from Zenodo
(https://doi.org/10.5281/zenodo.593938 and the CRAN repository (Sanchez-Reves et al.,
2022). DateLife's web application is maintained using RStudio's shiny server and the shiny
package open infrastructure, as well as Docker and OpenTree's infrastructure
(datesdatelife.opentreeoflife.org/datelife). datelife's R package stable version is available

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for installation stable version can be installed from the CRAN repository ()—using the

command install.packages(pkgs = "datelife") from within R. Development versions

are available from the DateLife's GitHub repository

(https://github.com/phylotastic/datelife) and can be installed using the command

devtools::install_github("phylotastic/datelife").
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#### SUPPLEMENTARY MATERIAL

Code used to generate all versions of this manuscript, the Supplementary Figures can 699 be viewed and downloaded from their Zenodo repository 700 (https://doi.org/10.5281/zenodo.6683667). Supplementary material, including code, 701 biological examples, benchmark results, data files and online-only appendices, can be 702 downloaded from the Dryad data repository (https://doi.org/10.5061/dryad.cnp5hqc6w), as 703 well as the benchmark of functionalities are available at datelifeMS1, datelife—examples, 704 and datelife\_benchmark repositories in LLSR's GitHub account in the Zenodo stable 705 repositories that host the reproducible manuscript 706 (https://doi.org/10.5281/zenodo.7435094), the biological examples 707 (https://doi.org/10.5281/zenodo.7435101), and the software benchmark (https://doi.org/10.5281/zenodo.7435106). Development versions corresponding to all of the above are hosted on GitHub, accesible at https://github.com/LunaSare/datelifeMS1, https://github.com/LunaSare/datelife examples, and 711 https://github.com/LunaSare/datelife benchmark.

713 Funding

Funding was provided by the US National Science Foundation (NSF) grants

ABI-1458603 to the DateLife project; DBI-0905606 to the National Evolutionary

Synthesis Center (NESCent), ABI-1458572 to the Phylotastic project, and ABI-1759846

to the Open Tree of Life project.

#### ACKNOWLEDGEMENTS

We thank Isabel Sanmartín, Daniele Silvestro, Rutger Vos and an anonymous 719 reviewer, for comments that greatly improved this manuscript. The DateLife project was born as a prototype tool aiming to provide these services the services describe in this paper, and was initially developed over a series of hackathons at the National Evolutionary Synthesis Center, NC, USA (Stoltzfus et al., 2013). We thank colleagues from the O'Meara 723 Lab at the University of Tennesse Tennessee Knoxville for suggestions, discussions and 724 software testing. The late National Evolutionary Synthesis Center (NESCent), which 725 sponsored hackathons that led to initial work on this project. The team that assembled 726 DateLife's first proof of concept: Tracy Heath, Jonathan Eastman, Peter Midford, Joseph 727 Brown, Matt Pennell, Mike Alfaro, and Luke Harmon. The Open Tree of Life project that 728 provides the open, metadata rich repository of trees used to construct DateLife's chronogram 729 database. The many scientists who publish their chronograms in an open, reusable form, and 730 the scientists who curate them for deposition in the Open Tree of Life repository. The NSF 731 for funding nearly all the above, in addition to the ABI grant that funded this project itself. 732

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