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

Abstract

Chronograms –phylogenies with branch lengths proportional to time– represent key 19 data on timing of evolutionary events for the study of natural processes in many areas of biological research. Chronograms also provide valuable information that can be used for education, science communication, and conservation policy decisions. Yet, achieving a high-quality reconstruction of a chronogram is a difficult and resource-consuming task. Here we present DateLife, a service-phylogenetic software implemented as an R package and an R Shiny web application available at www.datelife.org, that provides services for efficient and 25 easy discovery, summary, reuse, and reanalysis of node age data mined from a curated 26 database of expert, peer-reviewed, and openly available chronograms. The main DateLife 27 workflow starts with one or more scientific taxon names provided by a user. Names are 28 processed and standardized to a unified taxonomy, allowing DateLife to run a name match 29 across its local chronogram database that is curated from Open Tree of Life's phylogenetic repository, and extract all chronograms that contain at least two queried taxon names, along 31 with their metadata. Finally, node ages from matching chronograms are mapped using the 32 congruification algorithm to corresponding nodes from a chosen tree topologyusing the congruification algorithmon a tree topology, either extracted from Open Tree of Life's 34 synthetic phylogeny or one provided by the user. Congruified node ages are used as 35 secondary calibrations to date the chosen topology, with 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 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 41 ages occurring around topologically deeper nodes. Results Because the results from any

- software for scientific 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. We encourage the use of DateLife to help DateLife can help to increase 45
- awareness of the existing disparities among alternative hypotheses of dates for the same
- diversification events, and to support exploration of the effect of alternative chronogram 47
- . downst.

 m of evolutiona.

 /words: Tree; Phylogeny; S.

 fication; Supertree; Calibrations; .

 Word count: 6707-6817 hypotheses on downstream analyses, providing a framework for a more informed
- interpretation of evolutionary results.
- Keywords: Tree; Phylogeny; Scaling; Dating; Ages; Divergence times; Open Science; 50
- Congruification; Supertree; Calibrations; Secondary calibrations. 51
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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 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 a full dated tree. All of this entails specialized biological training,
taxonomic domain knowledge, and a significant amount of research time, computational
resources and funding.

Here we present the DateLife project which has the main goal of extracting and
exposing age data from published chronograms, making age data readily accessible to a
wider community for reuse and reanalysis in research, teaching, 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 chronogram database (McTavish et al., 2015) that stores data in a
computer-readable format (Vos et al., 2012); automated and programmatic ways of accessing
and downloading the data, also in a computer-readable format (Stoltzfus et al., 2013); and
methods to summarize and compare the data.

Description

DateLife's core software applications are implemented in the R package datelife, and

relies on functionalities from other biological R packages: ape (Paradis, Claude, & Strimmer, 2004), bold (Chamberlain, 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, Brown, & Winter, 2016), and taxize (Chamberlain, 2018; Chamberlain & Szöcs, 2013). Figure 1 provides a graphical summary of the There are three main steps of to the DateLife workflow: 1) creating a search query, 2) searching a database, and 3) summarizing results from the search.

Creating a Search Query

DateLife starts by processing an input consisting of 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" 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 processes input scientific names using a Taxonomic Name Resolution Service 93 (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, 95 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 homonym. DateLife implements TNRS using the Open Tree of Life (OpenTree) unified Taxonomy (OTT, Open Tree Of Life et al., 2016; Rees & Cranston, 2017) as standard, storing 100 taxonomic identification numbers (OTT ids) for further processing and analysis. Other 101 taxonomies currently supported by DateLife are the National Center of Biotechnology 102 Information (NCBI) taxonomic database (Schoch et al., 2020), the Global Biodiversity 103

Information Facility (GBIF) taxonomic backbone (GBIF Secretariat, 2022), and the Interim
Register of Marine and Non-marine Genera (IRMNG) database (Rees et al., 2017).

Besides binomial species names, DateLife accepts scientific names from any inclusive 106 taxonomic group (e.g., genus, family, tribe), as well as subspecific taxonomic variants (e.g., 107 subspecies, variants, strains). If a taxon name belongs to an inclusive taxonomic group, 108 DateLife has two alternative behaviors defined by the "get species from taxon" flag. If the 109 flag is active, DateLife retrieves all species names within a taxonomic group provided, from a 110 standard taxonomy of choice, and adds them to the search query. In this case, subspecific 111 variants are excluded. If the flag is inactive, DateLife excludes inclusive taxon names from 112 the search query, and species and subspecific variant names are processed as provided by the 113 user. The processed taxon names are saved as an R object of a newly defined class, 114 datelifeQuery, that is used in the following steps. This object contains the input names 115 standardized to a taxonomy of choice (OTT by default), the corresponding OTT id numbers, 116 and the topology of an input tree, if one was provided.

Searching a Chronogram Database

At the time of writing of this manuscript (Jun 11 Nov 07, 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,

state-of-the-art chronograms any time, through OpenTree's curator application

(https://tree.opentreeoflife.org/curator). As chronograms are added to Phylesystem, they

can be incorporated into the chronogram database of the datelife R package, which is

currently manually updated as new chronogram data is added to Phylesystem. The updated

database is assigned a new version number, followed by a package release on CRAN. Users

can directly run datelife functions to trigger an update of their local chronogram database,

to incorporate any new chronograms to their DateLife analysis before an official database

update is released on CRAN.

A DateLife search is implemented by matching processed taxon names provided by the 135 user to tip labels in the chronogram database. Chronograms with at least two matching 136 taxon names on their tip labels are identified and pruned down to preserve only the matched 137 taxa. These matching pruned chronograms are referred to as source chronograms. Total 138 distance in units of million years (Myr) between taxon pairs within each source chronogram 139 are stored as a patristic distance matrix (Fig. 1). The matrix format speeds up extraction of pairwise taxon ages of any queried taxa, as opposed to searching the ancestor node of a 141 pair of taxa in a "phylo" object or newick string. Finally, the patristic matrices are associated to the study citation where the original chronogram was published, and stored as an R object of the newly defined class datelifeResult.

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 isincludes:

1. The matching pruned chronograms as newick strings or "phylo" objects.

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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 within the group of interest.

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- 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.

Choosing a Topology

DateLife requires a tree topology to summarize age data upon. We recommend that 160 users provide as input a tree topology from the literature, or one of their own making. If no 161 topology is provided, DateLife automatically extracts one from the OpenTree synthetic tree, 162 a phylogeny currently encompassing 2.3 million taxa across all life, assembled from 1,239 163 published phylogenetic trees and OpenTree's unified Taxonomy, OTT (Open Tree Of Life et 164 al., 2019). Alternatively, DateLife can combine topologies from source chronograms using a 165 supertree approach (Criscuolo, Berry, Douzery, & Gascuel, 2006). To do this, DateLife first 166 identifies the source chronograms that form a grove, roughly, a sufficiently overlapping set of 167 taxa between trees, by implementing definition 2.8 for n-overlap from Ané et al. (2009). If 168 the source chronograms do not form a grove, the supertree reconstruction will fail. In rare cases, a group of trees can have multiple groves. By default, DateLife chooses the grove with 170 the most taxa, however, the "criterion = trees" flag allows the user to choose the grove with the most trees instead. The result is a single summary (or-i.e., supertree) topology, that 172 combines topologies from source chronograms in a grove.

Applying Secondary Calibrations

Once a topology is chosen, DateLife applies the congruification method (Eastman,
Harmon, & Tank, 2013) that find nodes belonging to the same clade across source
chronograms, and then extracts the corresponding node ages from patristic distance matrices
stored as a datelifeResult object. Note that by definition, these matrices store total

distance (time from tip to tip), assuming that the terminal taxa are coeval and occur at the
present. Hence, node ages correspond to half the values stored in the datelifeResult
matrices. A table of congruified node ages that can be used as calibrations for a dating
analysis is stored as a congruifiedCalibrations object.

For each congruent node, the pairwise distances that traverse that node are summarized 183 into a single summary matrix using classic summary statistics (i.e., mean, median, minimum 184 and maximum ages), and the Supermatrix Distance Method (SDM; Criscuolo et al., 2006), 185 which deforms patristic distance matrices by minimizing variance and then averaging them. 186 These single summary taxon pair age matrices are stored as summarized calibrations that 187 can be used as secondary calibrations to date a tree topology - with or without initial branch 188 lengths, using phylogenetic dating methods currently supported within DateLife: BLADJ 189 (Webb, Ackerly, & Kembel, 2008; Webb & Donoghue, 2005), MrBayes (Huelsenbeck & 190 Ronquist, 2001; Ronquist & Huelsenbeck, 2003), PATHd8 (Britton, Anderson, Jacquet, 191 Lundqvist, & Bremer, 2007), and treePL (Smith & O'Meara, 2012). 192

Dating a Tree Topology

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Dating a tree without branch lengths.—When producing or obtaining a tree 194 with branch lengths for a group of interest is not possible, DateLife can date a topology 195 without branch lengths, obtained from OpenTree or by implementing the supertree 196 approach described above, by applying the To date a tree topology when initial branch 197 lengths are unavailable, DateLife implements the Branch Length Adjuster (BLADJ-) 198 algorithm (Webb et al., 2008; Webb & Donoghue, 2005) algorithm, which requires no initial branch lengths. The algorithm starts by fixing, which only requires a tree topology with no branch lengths and at least two node ages to use as calibrations, one for the tree root 201 and one for any internal node of the topology. The BLADJ algorithm fixes ages for nodes 202 with calibration data upon the given tree topology. Then, it distributes time for assigns ages 203 to nodes with no data available age information by distributing time evenly between 204

calibrated nodes, minimizing age variance in the resulting chronogram. This approach has 205 proven useful for ecological analyses that require a phylogenetic time context (Webb et al., 206 2008). When there is conflict between ages of calibrated nodes, BLADJ ignores node ages 207 that are older than the age of a parent node. The BLADJ algorithm requires a root age 208 estimate to run. If Users can provide an appropriate root age estimate of their own or one 200 obtained from the literature. If a root age is not provided and there is no information on 210 the age of the root in the chronogram database, users can provide an estimate from the 211 literature. If none is provided, DateLife will choose an DateLife chooses a random age for 212 the rootso that it can return a dated topology. It will also, so that a dated tree topology 213 can be generated with BLADJ. In this case, DateLife will provide a conspicuous warning 214 message, so that users are aware that the root of the chronogram does not have information 215 available was chosen at random because there was no information available for it in the 216 chronogram database, along with suggestions on how the user can find and provide an 217 appropriate age for the root of the initial topology.

In the absence of genetic data, BLADJ is a very agnostic way to assign ages to nodes 219 with no available data, as it does not require any assumptions on the underlying model of 220 branch length distribution. It is however An alternative to BLADJ to date tree topologies 221 in the absence of initial branch lengths that is common practice in the literature is to use a 222 birth-death model to assign ages to nodes with no genetic data draw branch lengths (Jetz, 223 Thomas, Joy, Hartmann, & Mooers, 2012; Rabosky et al., 2018; Smith & Brown, 2018). To 224 do so, DateLife implements In addition to the initial tree topology and nodes with age 225 data, these methods require initial values of speciation and extinction rate parameters provided by the user. DateLife implements this approach with MrBayes (Huelsenbeck & 227 Ronquist, 2001; Ronquist & Huelsenbeck, 2003), using nodes with published age data as calibration priors on a fixed topology nodes of a tree topology with no branch lengths, a 229 simple birth-death model with parameters that can be determined speciation and extinction 230 rate parameters that are provided by the user, and no genetic data. However, BLADJ is the 231

default option in DateLife, as it does not require any information on diversification rates
for the phylogenetic sample to draw from a branch length distribution.

Dating a tree with branch lengths. – Relative branch lengths can provide key 234 information for phylogenetic dating, specifically for nodes without any calibration data 235 available. While using initial branch length data is the golden standard for phylogenetic 236 dating analyses, estimating trees with branch lengths proportional to substitution rates per 237 site requires obtaining primary data, assembling and curating a homology (orthology) 238 hypothesis, and choosing and implementing a method for phylogenetic inference. DateLife implements a workflow to streamline this process by applying open data from the Barcode of Life Data System, BOLD (Ratnasingham & Hebert, 2007) to obtain genetic markers for input taxa. By default, BOLD genetic sequences are aligned with MUSCLE (Edgar, 2004) using functions from the msa R package (Bodenhofer et al., 2015). Alternatively, sequences 243 can be aligned with MAFFT (Katoh, Asimenos, & Toh, 2009), using functions from the ape 244 R package (Paradis et al., 2004). The BOLD sequence alignment is then used to obtain 245 initial branch lengths with the accelerated transformation (ACCTRAN) parsimony 246 algorithm, which resolves ambiguous character optimization by assigning changes along 247 branches of the tree as close to the root as possible (Agnarsson & Miller, 2008), resulting in 248 older internal nodes as compared to other parsimony algorithms (Forest et al., 2005). The 249 parsimony branch lengths are then optimized using Maximum Likelihood, given the 250 alignment, the topology and a simple Jukes-Cantor model, producing a BOLD tree with 251 branch lengths proportional to expected number of substitutions per site. Both parsimony 252 and ML optimizations are done with functions from the phangorn package (Schliep, 2011). 253 Due to the computing load it requires, the BOLD workflow is currently only supported 254 through DateLife's R package. It is not yet available through the web application. 255

Phylogenetic dating methods supported in DateLife that incorporate branch length information from the input topology in combination with the secondary calibrations include:

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PATHd8, a non-clock, rate-smoothing method to date trees (Britton et al., 2007); treePL (Smith & O'Meara, 2012), a semi-parametric, rate-smoothing, penalized likelihood dating 259 method (Sanderson, 2002); and MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & 260 Huelsenbeck, 2003), a Bayesian inference program implementing Markov chain Monte Carlo 261 (MCMC) methods to estimate a posterior distribution of model parameters. 262

Visualizing Results

Finally, users can save all source and summary chronograms in formats allowing for 264 reuse and reanalysis, such as newick and the R "phylo" format. Input and summary 265 chronograms can be visualized and compared graphically, and users can construct their own 266 graphs using DateLife's chronogram plot generation functions available from the R package 267 datelifeplot (Sanchez-Reyes & O'Meara, 2022). 268

BENCHMARK

R package datelife code speed was tested on an Apple iMac with one 3.4 GHz Intel 270 Core is processor. We registered variation in computing time of query processing and search through the database relative to number of queried taxon names. Query processing time 272 increases roughly linearly with number of input taxon names, and increases considerably if Taxonomic Name Resolution Service (TNRS) is activated. Up to ten thousand names can be processed and searched in less than 30 minutes with the most time consuming settings. Once 275 names have been processed as described in methods, a name search through the chronogram 276 database can be performed in less than a minute, even with a very large number of taxon names (Fig. $\frac{21}{1}$). 278

datelife's code performance was evaluated with a set of unit tests designed and 279 implemented with the R package testthat (R Core Team, 2018) that were run both locally 280 with the devtools package (R Core Team, 2018), and on a public server using the continuous 281 integration tool of GitHub actions (https://docs.github.com/en/actions). At present, unit 282

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 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 Figure 3.—2. As a
second example, we analysed 289 bird species in the family Fringillidae that are included in
the NCBI taxonomy. The summary chronogram resulting One clade from the full summary
chronogram result from the DateLife analysis is shown in Figure 5, and Figure 3. The full
chronogram for all 289 species and the 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 296 sample includes two species of cardinals: the black-thighed grosbeak – Pheucticus tibialis and the crimson-collared grosbeak – Rhodothraupis celaeno; three species of buntings: the yellowhammer – Emberiza citrinella, the pine bunting – Emberiza leucocephalos and the 299 yellow-throated bunting – *Emberiza elegans*; and one species of tanager, the vegetarian finch – 300 Platyspiza crassirostris. Processing of input names found that Emberiza elegans is synonym 301 for Schoeniclus elegans in the default reference taxonomy (OTT v3.3, June 1, 2021). For a 302 detailed discussion on the state of the synonym, refer to Avibase (Avibase, 2022; Lepage, 303 2004; Lepage, Vaidya, & Guralnick, 2014). Discovering this synonym allowed assigning five 304 age data points for the parent node of Emberiza elegans, shown as Schoeniclus elegans in 305 Figure 3a2, which would not have had any data otherwise. 306

Searching the database. – DateLife used the processed input names to search the

local chronogram database and found 9 matching chronograms from 6 different studies (Fig. 3b2c). Three studies matched five input names (Barker, Burns, Klicka, Lanyon, & Lovette, 2015; Hedges, Marin, Suleski, Paymer, & Kumar, 2015; Jetz 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 315 topology for these taxa (Fig. 3e2d), and congruified and mapped age data to nodes in this 316 chosen topology, shown in Table 1. The name processing step allowed including five data 317 points for node "n4" (parent of Schoeniclus elegans; Fig. 3A) that would not have had any 318 data otherwise due to name mismatch. Age summary statistics per node were calculated 319 (Table 2) and used as calibrations to date the tree topology using the BLADJ algorithm. As 320 expected, more inclusive nodes (e.g., node "n1") have more variance in age data than less 321 inclusive nodes (e.g., node "n5"). Median summary age data for node "n2" was excluded as 322 final calibration because it is older than the median age of a more inclusive node, "n1" (Fig. 323 3c42g). 324

An Example with the Family of True Finches

Creating a query. To obtain ages for all species within the family of true finches, 326 Fringillidae, we ran a DateLife query using the "get species from taxon" flag, which gets all 327 recognized species names within a named group from a taxonomy of choice. Following the 328 NCBI taxonomy, our DateLife query has 289 Fringillidae species names. This 329 taxon-constrained approach implies that the full DateLife analysis will be performed using a 330 tree topology and ages available for species names from a given taxonomic group, which do 331 not necessarily correspond to a monophyletic group. Users can change this behavior by 332 providing all species names corresponding to a monophyletic group as input for a DateLife 333 search, or a monophyletic tree to construct a DateLife summary. 334

Searching the database.— Next, we used the processed species names in our

DateLife query to identify chronograms with at least two Fringillidae species as tip taxa.

The DateLife search identified 19 chronograms matching this criteria, published in 13

different studies (Barker et al., 2013, 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 al., 2014; Roquet, Lavergne, & Thuiller, 2014; Uyeda,

Pennell, Miller, Maia, & McClain, 2017). Once identified, DateLife pruned these matching

chronograms to remove tips that do not belong to the queried taxon names, and transformed

these pruned chronograms to pairwise distance matrices, revealing 1,206 different age data

points available for species within the Fringillidae (Supplementary Table S1).

Summarizing search results.— The final step 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" section, a tree topology to summarize age data upon is required. By
default, DateLife uses the topology from OpenTree's synthetic tree that contains all taxa
from the search query. According to OpenTree's synthetic tree, species belonging to the

family Fringillidae do not form a monophyletic group (Fig. 4aSupplementary Fig. S1).

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. 4bSupplementary Fig. S2; Open Tree Of Life

et al., 2019).

Source All 19 source chronograms (Supplementary Figs. S2-S20S5-S23) were 355 congruified to OpenTree's topology shown in Figure 4bSupplementary Figure S2, reducing 356 the original 1,206 node age data set to 818 different data points (Supplementary Table S2) 357 that can could be used as calibrations for the chosen topology (Fig. 4bSupplementary Fig. 358 S2). The congruent node age data points were summarized for each node, resulting in 194 359 summary node ages. From these 21 were excluded as secondary calibrations because they 360 were older than the ancestral node. The remaining 173 summary node ages were used as 361 secondary calibrations to obtain a fully dated (and resolved) phylogeny with the program 362 BLADJ (Fig. 5). Supplementary Figure S3). Results for a subgroup are shown in Figure 3. 363

CROSS-VALIDATION TEST

We performed a cross validation test of a DateLife analysis using the Fringillidae 365 source chronograms obtained above (Supplementary Figs. S2-S20). We used as S5-S23). As 366 inputs for a DateLife analysis, we used all individual tree topologies from each of the 19 367 source chronograms from 13 studies, treating their node ages as unknown. We congruified 368 node ages extracted from chronograms from all other studies upon the individual topologies, 369 effectively excluding original ages from each topology. Finally, average node ages per node 370 were applied as secondary calibrations and smoothed with the BLADJ algorithm. We found that node ages from the original studies, and ages estimated using all other age data 372 available are largely generally correlated (Fig. 64). For five studies, DateLife tended to 373 underestimate ages for topologically deeper nodes (those with many descendant taxa, aka 374 "closer to the root") relative to the original estimate, and overestimate ages for nodes closer 375 to the tips. Accordingly, root ages are generally older in the original study than estimated 376

using cross-validated ages (Supplementary Fig. \$154). In general, topologically deeper nodes
display the largest age variation between node ages from the original chronograms and ages
summarized with DateLife.

380 DISCUSSION

DateLife's goal is to improve availability and accessibility of state-of-the-art data on
evolutionary time frame of organisms, to allow users from all areas of science and with all
levels of expertise 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 –besides the scientific names of the species or group users want to work
with, to use any of its functionalities.

A total of 99,474 unique terminal taxa are represented in DateLife's database. 387 Incorporation of more chronograms into the database will continue to improve DateLife's 388 services. One option to increase the number of chronograms in the DateLife database is the 380 Dryad data repository. Methods to automatically mine chronograms from Dryad could be 390 designed and implemented. However, Dryad's metadata system has no information to 391 automatically detect branch length units, and those would still need to be determined 392 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 beyond, we emphasize that DateLife's results —as well as any insights gleaned from them, largely depend on the quality of the source chronograms: low quality chronograms will produce low quality results. The "garbage in, garbage out" problem has long been recognised in supertree methods for summarizing phylogenetic trees (Bininda-Emonds et al., 2004). We

note that this is a surfacing issue of any automated tool for biological data analysis. For
example, DNA riddled with sequencing errors will produce generally poor alignments that
will return biased evolutionary hypothesis, independently of the quality of the analysis
software used. Again, we urge readers and DateLife users to explore all input chronograms
before using a summary chronogram resulting from a DateLife workflow.

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

Conflict in estimated ages among alternative studies is common in the literature. See, 415 for example, the robust ongoing debate about crown group age of angiosperms 416 (Barba-Montoya, Reis, Schneider, Donoghue, & Yang, 2018; Magallón, Gómez-Acevedo, 417 Sánchez-Reyes, & Hernández-Hernández, 2015; Ramshaw et al., 1972; Sanderson & Doyle, 418 2001; Sauquet, Ramírez-Barahona, & Magallón, 2021). Alternative source chronograms 419 available for the same taxa have potentially been estimated implementing different types of 420 calibrations, which affects the resulting node age estimates. For example, in the DateLife 421 analysis of the Fringillidae shown above, the chronograms from one study (Burns et al., 2014) were inferred using molecular substitution rate estimates across birds (Weir & Schluter, 2008), and have much older age estimates for the same nodes than chronograms that were inferred using fossil calibrations (Figs. 5, 6Fig. 4; Supplementary Figs. S1c, 425 \$4\$\frac{54}{6}\$. Another source of conflict in estimated node ages can arise from different 426 placements for the same calibration, which would imply fundamentally distinct evolutionary

hypotheses (Antonelli et al., 2017). For example, two independent researchers working on the same clade should both carefully select and justify their choices of fossil calibration 420 placement. Yet, if one researcher concludes that a fossil should calibrate the ingroup of a 430 clade, while another researcher concludes that the same fossil should calibrate the outgroup 431 of the clade, the resulting age estimates will differ, as the placement of calibrations as stem 432 or crown group is known to significantly affect estimates of time of lineage divergence 433 (Sauguet, 2013). Finally, placement of calibrations also affects uncertainty of node age 434 estimates. For example, nodes that are sandwiched between a calibrated node and a 435 calibrated root have less freedom of movement and hence narrower confidence intervals (Vos 436 & Mooers, 2004), which inflates precision for nodes without calibrations but does not 437 necessarily improve accuracy of the estimated ages. 438

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

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DateLife constructs summary chronograms using node ages extracted from existing
chronograms, i.e. secondary calibrations. In general, the scientific community has more
confidence in chronograms using primary calibrations, where the dated tree is generated from
a single analysis where carefully chosen fossil calibrations are the source of absolute time
information, than in analyses dated using secondary calibrations (Antonelli et al., 2017;
Garzón-Orduña, Silva-Brandão, Willmott, Freitas, & Brower, 2015; Graur & Martin, 2004;
Sauquet, 2013; Sauquet et al., 2012; Schenk, 2016; Shaul & Graur, 2002). However,
implementation of primary calibrations is difficult: it requires specialized expertise and

training to discover, place and apply calibrations appropriately (Hipsley & Müller, 2014; 454 Ksepka et al., 2011). One approach is to use fossils that have been widely discussed and 455 previously curated as calibrations to date other trees (Ksepka et al., 2011; Sauquet, 2013), 456 and making sure that all data reflect a coherent evolutionary history (Sauquet, 2013), as for 457 example done by Antonelli et al. (2017). The Fossil Calibration Database provides data for 458 220 primary calibration points encompassing flowering plants and metazoans, that have been 450 curated by experts and used for dating analysis in peer-reviewed publications (Ksepka et al., 460 2015). This database facilitates the use of expert primary fossil calibrations in new 461 phylogenetic dating analyses. Yet, users still require the expertise to locate and calibrate 462 appropriate nodes in their phylogenies which correspond with fossils available in the 463 database.

Recently, Powell, Waskin, and Battistuzzi (2020) showed in a simulation study that 465 secondary calibrations using node ages based on previous molecular clock analyses can be as 466 good as primary calibrations. Using several secondary calibrations (as opposed to just one) 467 can provide sufficient information to alleviate or even neutralize potential biases (Graur & 468 Martin, 2004; Sauguet, 2013; Shaul & Graur, 2002). Our cross validation analysis also 460 provides insight into the application of secondary calibrations. Node ages summarized with 470 DateLife and those from the original studies are well correlated (Supplementary Figs. 471 \$2-\$20\$5-\$23). We also note that DateLife estimates for nodes closer to the root tend to be 472 slightly younger than ages from the original studies. In contrast, nodes closer to the tips 473 tend to be slightly older when estimated using our secondary calibrations than ages from the 474 original studies. The only exception to this trend was observed in Burns et al. (2014) chronogram, which generally displays much younger node ages when estimated using 476 secondary calibrations than the original study (Supplementary Figs. \$1, \$5\$4c, \$10), supporting previous observations (Sauguet et al., 2012; Schenk, 2016). However, these younger dates are more likely an example of how multiple secondary calibrations can correct 470 erroneous estimates, as dates on the Burns et al. (2014) tree were obtained using a single 480

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secondary calibration based on a previously estimated molecular evolution rate across birds from Weir and Schluter (2008), and appear as major outliers compared to alternate estimates for the same nodes based on primary fossil calibrations (Fig. 53, Supplementary Fig. S3).

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).

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.

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

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 chronograms allows other researchers to do such analyses themselves reproducibly, and without needing permission. Though, of course, they should follow proper citation practices,

especially for the source chronogram studies.

The exercise of summarizing age data from across multiple studies provides the 507 opportunity to work with a chronogram that reflects a unified evolutionary history for a 508 lineage, by putting together evidence from different hypotheses. The largest, and 509 taxonomically broadest chronogram currently available from OpenTree was constructed 510 summarizing age data from 2,274 published chronograms using NCBI's taxonomic tree as 511 backbone (Hedges et al., 2015). A summarizing exercise may also amplify the effect of 512 uncertainty and errors in source data, and blur parts of the evolutionary history of a lineage 513 that might only be reflected in source chronograms and lost on the summary chronogram 514 (Sauguet et al., 2021). 515

Effects of Taxon Sampling on Downstream Analyses

Analysis of species diversification of simulated and empirical phylogenies suggest that
using a more completely sampled phylogeny provides estimates that are closer to the true
diversification history than when analysing incompletely sampled phylogenies (Chang,
Rabosky, & Alfaro, 2020; Cusimano, Stadler, & Renner, 2012; Sun et al., 2020). Ideally,
phylogenies should be completed using genetic data, but this is a time-consuming and
difficult task to achieve for many biological groups. Hence, DateLife's workflow features
different ways of assigning divergence times to taxa with missing the absence of branch
length data and calibrations and branch lengths for certain taxa.

Completing a phylogeny using a stochastic birth-death polytomy resolver and a
backbone taxonomy is a common practice in scientific publications: Jetz et al. (2012),
created a chronogram of all 9,993 bird species, where 67% had molecular data and the rest
was simulated; Rabosky et al. (2018) created a chronogram of 31,536 ray-finned fishes, of
which only 37% had molecular data; Smith and Brown (2018) constructed a chronogram of
353,185 seed plants where only 23% had molecular data. These stochastically resolved

chronograms return diversification rates estimates that appear less biased than those
estimated from their incompletely sampled counterparts, even with methods that account for
missing lineages by using sampling fractions (Chang et al., 2020; Cusimano et al., 2012), but
can also introduce spurious patterns of early bursts of diversification (Cusimano & Renner,
2010; Sun et al., 2020).

Taxonomy-based stochastic polytomy resolvers also introduce topological differences in 536 phylogenetic trees. The study of macroevolutionary processes largely depends on an 537 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 chronograms from the same biological group can infer very different species diversification patterns (Rabosky, 2015; Title & Rabosky, 2016). Similarly, there are worries that patterns of morphological evolution 542 cannot be accurately inferred with phylogenies that have been resolved stochastically over a 543 taxonomic backbone, as any patterns would be erased by randomization (Rabosky, 2015). We note that the same applies for geography- and morphology-dependent diversification 545 analysis. Hence, we suggest that phylogenies that have been processed with taxonomy-based 546 stochastic polytomy resolvers, including certain summary chronograms from a DateLife 547 analysis, can be useful as null or neutral models, representing the case of a diversification 548 process that is independent of traits and geographical scenario. 549

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 generate a
fully resolved, calibrated tree of all modern and extinct taxa using a single taxonomy, a
single calibration, 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

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consistency of DateLife's results with prior work.

Conclusions 558

Knowledge of the evolutionary time frame of organisms is key to many research areas: 559 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 publicly funded research have resulted in vast amounts of chronograms that are already available in scientific publications, but functionally hidden from the public and scientific community for reuse. 566

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

Availability

The DateLife software is free and open source. It can be used online through its R shiny web application hosted at http://www.datelife.org, and locally through the datelife R package, available from Zenodo (https://doi.org/10.5281/zenodo.593938 and the CRAN

repository (Sanchez-Reyes 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 (datelife.opentreeoflife.org). datelife's stable version can be
installed from the CRAN repository using the command install.packages(pkgs =
"datelife") from within R. Development versions are available from DateLife's GitHub
repository (https://github.com/phylotastic/datelife) and can be installed using the
command devtools::install_github("phylotastic/datelife").

Supplementary Material

Supplementary Figures can be viewed and downloaded from their Zenodo repository 590 (https://doi.org/10.5281/zenodo.6683667). Supplementary material, including code, 591 biological examples, benchmark results, data files and online-only appendices, can be 592 downloaded from the Dryad data repository (https://doi.org/10.5061/dryad.cnp5hqc6w), as 593 well as in the Zenodo stable repositories that host the reproducible manuscript 594 (https://doi.org/10.5281/zenodo.7435094), the biological examples 595 (https://doi.org/10.5281/zenodo.7435101), and the software benchmark 596 (https://doi.org/10.5281/zenodo.7435106). Development versions corresponding to all of the 597 above are hosted on GitHub, accesible at https://github.com/LunaSare/datelifeMS1, 598 https://github.com/LunaSare/datelife_examples, and 599 https://github.com/LunaSare/datelife_benchmark. 600

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	Node Name	taxon A	taxon B Systematic Biology	Node Age	Study chronogram Page 42 of 48
1	n1	Emberiza citrinella	Platyspiza crassirostris	9.6509881	Jetz et al. 2012 – chronogram 1
2	n1	Pheucticus tibialis	Schoeniclus elegans	14.3336520	Barker et al. 2015 – chronogram 1
3	n1	Rhodothraupis celaeno	Schoeniclus elegans	16.2984859	Barker et al. 2015 – chronogram 2
4	n1	Emberiza citrinella	Platyspiza crassirostris	16.9499615	Hooper et al. 2017
5	n1	Rhodothraupis celaeno	Schoeniclus elegans	21.5140867	Jetz et al. 2012 – chronogram 2
6	n1	Emberiza leucocephalos	Platyspiza crassirostris	22.0000000	Hedges et al. 2015 – chronogram 2
7	n1	Emberiza citrinella	Platyspiza crassirostris	22.3757277	Hedges et al. 2015 – chronogram 1
8	n2	Platyspiza crassirostris	Rhodothraupis celaeno	7.9691925	Jetz et al. 2012 - chronogram 1
9	n2	Platyspiza crassirostris	Rhodothraupis celaeno	19.7085830	Jetz et al. 2012 - chronogram 2
10	n2	Platyspiza crassirostris	Rhodothraupis celaeno	19.7085900	Hedges et al. 2015 – chronogram 2
11	n2	Platyspiza crassirostris	Rhodothraupis celaeno	19.7128363	Hedges et al. 2015 – chronogram 1
12	n2	Pheucticus tibialis	Platyspiza crassirostris	44.2958603	Burns et al. 2014
13	n3	Pheucticus tibialis	Rhodothraupis celaeno	10.5304440	Barker et al. 2015 - chronogram 1
14	n3	Pheucticus tibialis	Rhodothraupis celaeno	10.5379092	Barker et al. 2013
15	n3	Pheucticus tibialis	Rhodothraupis celaeno	11.2095375	Barker et al. 2015 – chronogram 2
16	n3	Pheucticus tibialis	Rhodothraupis celaeno	18.1570685	Hedges et al. 2015 – chronogram 1
17	n4	Emberiza citrinella	Schoeniclus elegans	5.3238969	Jetz et al. 2012 - chronogram 1
18	n4	Emberiza leucocephalos	Schoeniclus elegans	9.8622460	Barker et al. 2015 – chronogram 1
19	n4	Emberiza leucocephalos	Schoeniclus elegans	10.3391445	Jetz et al. 2012 – chronogram 2
20	n4	Emberiza leucocephalos	Schoeniclus elegans	11.7317630	Barker et al. 2015 – chronogram 2
21	n4	Emberiza citrinella	Schoeniclus elegans	12.5133870	Hooper et al. 2017
22	n5	Emberiza citrinella	Emberiza leucocephalos	0.1407015	Jetz et al. 2012 - chronogram 1
23	n5	Emberiza citrinella	Emberiza leucocephalos	0.1516230	Hedges et al. 2015 – chronogram 2
24	n5	Emberiza citrinella	Emberiza leucocephalos	0.2011990	Barker et al. 2015 – chronogram 1
25	n5	Emberiza citrinella	Emberiza leucocephalos	0.2409300	Barker et al. 2015 – chronogram 2
26	n5	Emberiza citrinella	Emberiza leucocephalos	0.2732460	Jetz et al. 2012 – chronogram 2
27	n5	Emberiza citrinella	http://mc.manuscriptcentral.com/ Emberiza leucocephalos	o.5760260	Hedges et al. 2015 – chronogram 1
28	n5	Emberiza citrinella	Emberiza leucocephalos	2.2898230	Hooper et al. 2017

Node.Name	Min.Age	Q1	Median.Age	Mean.Age	Q3	Max.Age	Variance	SD
n1	9.6509881	15.316069	16.94996	17.5889860	21.757043	22.375728	22.2431847	4.7162681
n2	7.9691925	19.708583	19.70859	22.2790124	19.712836	44.295860	177.3279940	13.3164558
n3	10.5304440	10.536043	10.87372	12.6087398	12.946420	18.157069	13.7831237	3.7125630
n4	5.3238969	9.862246	10.33914	9.9540875	11.731763	12.513387	7.8263782	2.7975665
n5	0.1407015	0.176411	0.24093	0.5533641	0.424636	2.289823	0.6079318	0.7796998

https://datadryad.org/stash/dataset/doi:10.5061/dryad.cnp5hqc6w









