

For peer review only. Do not cite.

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

Journal:	Systematic Biology
Manuscript ID	USYB-2022-152.R2
Manuscript Type:	Software for Systematics and Evolution
Date Submitted by the Author:	n/a
Complete List of Authors:	Sanchez Reyes, Luna Luisa; University of California System, Life and Environmental Sciences McTavish, Emily Jane; University of California System, Life and Environmental Sciences O'Meara, Brian; The University of Tennessee Knoxville, EEB
Keywords:	phylogeny, scaling, dating, ages, divergence times, open science, congruification, supertree, secondary calibrations, mrca
Note: The following files were submitted by the author for peer review, but cannot be converted to PDF. You must view these files (e.g. movies) online.	
systematic-biology.csl paper_references.bib	

SCHOLARONE™ Manuscripts

- DateLife: leveraging databases and analytical tools to reveal the dated Tree of Life
- Luna L. Sánchez Reyes^{1,2}, Emily Jane McTavish¹, & Brian O'Meara²
 - ¹ University of California, Merced, USA
- ² University of Tennessee, Knoxville, USA

- Department of Life and Environmental Sciences, University of California, Merced, CA 95343, USA.
- Department of Ecology and Evolutionary Biology, University of Tennessee, Knoxville,
- ⁹ 446 Hesler Biology Building, Knoxville, TN 37996, USA.
- The authors made the following contributions. Luna L. Sánchez Reyes: Data curation,
- 11 Investigation, Software, Visualization, Validation, Writing Original Draft Preparation,
- Writing Review & Editing; Emily Jane McTavish: Resources, Software, Writing Review &
- Editing; Brian O'Meara: Conceptualization, Funding acquisition, Methodology, Resources,
- Software, Supervision, Writing Review & Editing.
- Correspondence concerning this article should be addressed to Luna L. Sánchez Reyes, .
- E-mail: sanchez.reyes.luna@gmail.com

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 phylogenetic software implemented as an R package and an R Shiny web application available at www.datelife.org, that provides services for efficient and easy 25 discovery, summary, reuse, and reanalysis of node age data mined from a curated database of 26 expert, peer-reviewed, and openly available chronograms. The main DateLife workflow starts 27 with one or more scientific taxon names provided by a user. Names are processed and 28 standardized to a unified taxonomy, allowing DateLife to run a name match 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 with their 31 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 or without initial 35 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 ages occurring around topologically deeper nodes. 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. DateLife can help to increase awareness of the existing disparities
- among alternative hypotheses of dates for the same diversification events, and to support 45
- exploration of the effect of alternative chronogram hypotheses on downstream analyses,
- providing a framework for a more informed interpretation of evolutionary results. 47
- , Sc.
 .dibrations, Keywords: Tree; Phylogeny; Scaling; Dating; Ages; Divergence times; Open Science; 48
- Congruification; Supertree; Calibrations; Secondary calibrations.
- Word count: 7484 50

75

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). There are three main steps 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 90 (TNRS), which increases the probability of correctly finding the queried taxon names in the 91 chronogram database. TNRS detects, corrects and standardizes name misspellings and typos, variant spellings and authorities, and nomenclatural synonyms to a single taxonomic 93 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 taxonomic identification numbers (OTT ids) for further processing and analysis. Other taxonomies currently supported by DateLife are the National Center of Biotechnology Information (NCBI) taxonomic database (Schoch et al., 2020), the Global Biodiversity 100 Information Facility (GBIF) taxonomic backbone (GBIF Secretariat, 2022), and the Interim 101

Register of Marine and Non-marine Genera (IRMNG) database (Rees et al., 2017).

Besides binomial species names, DateLife accepts scientific names from any inclusive 103 taxonomic group (e.g., genus, family, tribe), as well as subspecific taxonomic variants (e.g., 104 subspecies, variants, strains). If a taxon name belongs to an inclusive taxonomic group, 105 DateLife has two alternative behaviors defined by the "get species from taxon" flag. If the 106 flag is active. DateLife retrieves all species names within a taxonomic group provided, from a 107 standard taxonomy of choice, and adds them to the search query. In this case, subspecific 108 variants are excluded. If the flag is inactive, DateLife excludes inclusive taxon names from 100 the search query, and species and subspecific variant names are processed as provided by the 110 user. The processed taxon names are saved as an R object of a newly defined class, 111 datelifeQuery, that is used in the following steps. This object contains the input names 112 standardized to a taxonomy of choice (OTT by default), the corresponding OTT id numbers, 113 and the topology of an input tree, if one was provided. 114

Searching a Chronogram Database

At the time of writing of this manuscript (Nov 21, 2023), DateLife's chronogram 116 database latest version consist of 253 chronograms published in 187 different studies, 117 encompassing close to 100 k (99 474) different species. It is curated from OpenTree's 118 phylogenetic database, the Phylogystem, an open database of expert and peer-reviewed 119 phylogenetic knowledge with rich metadata and a wide taxonomic scope (McTavish et al., 120 2015). We expect DateLife's database to largely overlap with OpenTree's phylogenetic 121 database taxonomic coverage, where Chordata and Embryophyta are nearly fully sampled. In contrast, Bacteria, Fungi, Nematoda, and Insecta, currently present a large gap between 123 the number of named species and what has phylogenetic information in OpenTree's synthetic tree. It is likely that users working with the former groups will get results from a DateLife 125 analysis. If none of the user's species are found, the software will indicate the lack of age 126 data for the queried taxa in the database.

144

147

150

A unique feature of the Phylesystem is that any user can add new published, 128 state-of-the-art chronograms any time, through OpenTree's curator application 129 (https://tree.opentreeoflife.org/curator). Relying on an open source database permits an 130 automatic and reproducible assembly of DateLife's chronogram database, which is stored and 131 navigable as an R data object within the datelife R package. As chronograms are added to 132 Phylesystem, they can be incorporated into the chronogram database within the datelife 133 R package, by manually triggering an update. The updated datelife database is assigned a 134 new version number, followed by a package release on CRAN. We encourage users to submit 135 published chronograms to OpenTree's phylogenetic database, so that their taxon of interest 136 can be included in future DateLife searches. Users can directly run datelife functions to 137 trigger an update of their local chronogram database, to incorporate any new chronograms 138 to their DateLife analysis before a datelife database update is released on CRAN.

A DateLife search is implemented by matching processed taxon names provided by the 140 user to tip labels in the chronogram database. Chronograms with at least two matching taxon names on their tip labels are identified and pruned down to preserve only the matched taxa. These matching pruned chronograms are referred to as source chronograms. Total distance in units of million years (Myr) between taxon pairs within each source chronogram are stored as a patristic distance matrix. The matrix format speeds up extraction of pairwise taxon ages of any queried taxa, as opposed to searching the ancestor node of a pair of taxa 146 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 148 the newly defined class datelifeResult. 149

Summarizing Search Results

Summary information is extracted from the datelifeResult object to inform 151 decisions for subsequent steps in the analysis workflow. Basic summary information available 152 to the user includes: 153

161

164

- 1. The matching pruned chronograms as newick strings or "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 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.

Choosing a Topology

DateLife requires a tree topology to summarize age data upon. We recommend that 165 users provide as input a tree topology from the literature, or one of their own making. If no 166 topology is provided, DateLife automatically extracts one from the OpenTree synthetic tree, 167 a phylogeny currently encompassing 2.3 million taxa across all life, assembled from 1,239 168 published phylogenetic trees and OpenTree's unified Taxonomy, OTT (Open Tree Of Life et 169 al., 2019). Alternatively, DateLife can combine topologies from source chronograms using a 170 supertree approach (Criscuolo, Berry, Douzery, & Gascuel, 2006). To do this, DateLife first 171 identifies the source chronograms that form a grove, roughly, a sufficiently overlapping set of 172 taxa between trees, by implementing definition 2.8 for n-overlap from Ané et al. (2009). If 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 175 the most taxa, however, the "criterion = trees" flag allows the user to choose the grove with 176 the most trees instead. The result is a single summary (i.e., supertree) topology, that 177 combines topologies from source chronograms in a grove. 178

Applying Secondary Calibrations

Once a topology is chosen, DateLife applies the congruification method (Eastman, 180 Harmon, & Tank, 2013) that find nodes belonging to the same clade across source 181 chronograms, and then extracts the corresponding node ages from patristic distance matrices 182 stored as a datelifeResult object. Note that by definition, these matrices store total 183 distance (time from tip to tip), assuming that the terminal taxa are coeval and occur at the 184 present. Hence, node ages correspond to half the values stored in the datelifeResult 185 matrices. A table of congruified node ages that can be used as calibrations for a dating 186 analysis is stored as a congruifiedCalibrations object. 187

For each congruent node, the pairwise distances that traverse that node are summarized 188 into a single summary matrix using classic summary statistics (i.e., mean, median, minimum 180 and maximum ages), and the Supermatrix Distance Method (SDM; Criscuolo et al., 2006), 190 which deforms patristic distance matrices by minimizing variance and then averaging them. 191 These single summary taxon pair age matrices are stored as summarized calibrations that 192 can be used as secondary calibrations to date a tree topology - with or without initial branch 193 lengths, using phylogenetic dating methods currently supported within DateLife: BLADJ (Webb, Ackerly, & Kembel, 2008; Webb & Donoghue, 2005), MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), PATHd8 (Britton, Anderson, Jacquet, 196 Lundqvist, & Bremer, 2007), and treePL (Smith & O'Meara, 2012).

Dating a Tree Topology

198

201

203

204

205

Dating a tree without branch lengths. To date a tree topology when initial 199 branch lengths are unavailable, DateLife implements the Branch Length Adjuster (BLADJ) algorithm (Webb et al., 2008; Webb & Donoghue, 2005), 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 and one for any internal node of the topology. The BLADJ algorithm fixes ages for nodes with calibration data upon the given tree topology. Then, it assigns ages to nodes with no available age information by distributing time evenly between calibrated nodes, minimizing

age variance in the resulting chronogram. This approach has proven useful for ecological 206 analyses that require a phylogenetic time context (Webb et al., 2008). When there is conflict 207 between ages of calibrated nodes, BLADJ ignores node ages that are older than the age of a 208 parent node. The BLADJ algorithm requires a root age to run. Users can provide an 209 appropriate root age estimate of their own or one obtained from the literature. If a root age 210 is not provided and there is no information on the age of the root in the chronogram 211 database, DateLife chooses a random age for the root, so that a dated tree topology can be 212 generated with BLADJ. In this case, DateLife will provide a conspicuous warning message, 213 so that users are aware that the root of the chronogram was chosen at random because there 214 was no information available for it in the chronogram database, along with suggestions on 215 how the user can find and provide an appropriate age for the root of the initial topology. 216

An alternative to BLADJ to date tree topologies in the absence of initial branch 217 lengths that is common practice in the literature is to use a birth-death model to draw 218 branch lengths (Jetz, Thomas, Joy, Hartmann, & Mooers, 2012; Rabosky et al., 2018; Smith 219 & Brown, 2018). In addition to the initial tree topology and nodes with age data, these 220 methods require initial values of speciation and extinction rate parameters provided by the 221 user. DateLife implements this approach with MrBayes (Huelsenbeck & Ronquist, 2001; 222 Ronquist & Huelsenbeck, 2003), using nodes with published age data as calibration priors on 223 nodes of a tree topology with no branch lengths, a simple birth-death model with speciation 224 and extinction rate parameters that are provided by the user, and no genetic data. However, 225 BLADJ is the default option in DateLife, as it does not require any information on 226 diversification rates for the phylogenetic sample to draw from a branch length distribution. 227

Dating a tree with branch lengths.—Relative branch lengths can provide key information for phylogenetic dating, specifically for nodes without any calibration data available. While using initial branch length data is the golden standard for phylogenetic dating analyses, obtaining such information from scratch is not an easy task: it requires

obtaining primary data, assembling and curating a homology (orthology) hypothesis, and 232 choosing and implementing a method for phylogenetic inference. DateLife implements a 233 workflow to streamline this process by applying open data from the Barcode of Life Data 234 System, BOLD (Ratnasingham & Hebert, 2007) to obtain genetic markers for input taxa. 235 By default, BOLD genetic sequences are aligned with MUSCLE (Edgar, 2004) using 236 functions from the msa R package (Bodenhofer et al., 2015). Alternatively, sequences can be 237 aligned with MAFFT (Katoh, Asimenos, & Toh, 2009), using functions from the ape R 238 package (Paradis et al., 2004). The BOLD sequence alignment is then used to obtain initial 239 branch lengths with the accelerated transformation (ACCTRAN) parsimony algorithm, 240 which resolves ambiguous character optimization by assigning changes along branches of the 241 tree as close to the root as possible (Agnarsson & Miller, 2008), resulting in older internal 242 nodes as compared to other parsimony algorithms (Forest et al., 2005). The parsimony branch lengths are then optimized using Maximum Likelihood, given the alignment, the topology and a simple Jukes-Cantor model, producing a BOLD tree with branch lengths proportional to expected number of substitutions per site. Both parsimony and ML optimizations are done with functions from the phangorn package (Schliep, 2011). Due to 247 the computing load it requires, the BOLD workflow is currently only supported through DateLife's R package. It is not yet available through the web application.

Phylogenetic dating methods supported in DateLife that incorporate branch length information from the input topology in combination with the secondary calibrations include:
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 method (Sanderson, 2002); and MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), a Bayesian inference program implementing Markov chain Monte Carlo (MCMC) methods to estimate a posterior distribution of model parameters.

273

274

275

276

277

278

279

280

281

282

Finally, users can save all source and summary chronograms in formats 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).

Benchmark

R package datelife code speed was tested on an Apple iMac with one 3.4 GHz Intel 264 Core is processor. We registered variation in computing time of query processing and search 265 through the database relative to number of queried taxon names. Query processing time 266 increases roughly linearly with number of input taxon names, and increases considerably if 267 Taxonomic Name Resolution Service (TNRS) is activated. Up to ten thousand names can be 268 processed and searched in less than 30 minutes with the most time consuming settings. Once 269 names have been processed as described in methods, a name search through the chronogram 270 database can be performed in less than a minute, even with a very large number of taxon 271 names (Fig. 1). 272

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

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 2. As a second example, we analysed 289 bird species in the family Fringillidae that are included in the NCBI taxonomy. One clade from the full summary chronogram result from the DateLife analysis is shown 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 289 sample includes two species of cardinals: the black-thighed grosbeak – Pheucticus tibialis 290 and the crimson-collared grosbeak - Rhodothraupis celaeno; three species of buntings: the 291 yellowhammer – Emberiza citrinella, the pine bunting – Emberiza leucocephalos and the 292 vellow-throated bunting – Emberiza elegans; and one species of tanager, the vegetarian finch – 293 Platyspiza crassirostris. Processing of input names found that Emberiza elegans is synonym 294 for Schoeniclus elegans in the default reference taxonomy (OTT v3.3, June 1, 2021). For a 295 detailed discussion on the state of the synonym, refer to Avibase (Avibase, 2022; Lepage, 296 2004; Lepage, Vaidya, & Guralnick, 2014). Discovering this synonym allowed assigning five 297 age data points for the parent node of Emberiza elegans, shown as Schoeniclus elegans in 298 Figure 2, 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 from 6 different studies (Fig. 2c). 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 308 topology for these taxa (Fig. 2d), and congruified and mapped age data to nodes in this 309 chosen topology, shown in Table 1. The name processing step allowed including five data 310 points for node "n4" (parent of Schoeniclus elegans) that would not have had any data 311 otherwise due to name mismatch. Age summary statistics per node were calculated (Table 2) 312 and used as calibrations to date the tree topology using the BLADJ algorithm. As expected, 313 more inclusive nodes (e.g., node "n1") have more variance in age data than less inclusive 314 the media nodes (e.g., node "n5"). Median summary age data for node "n2" was excluded as final 315 calibration because it is older than the median age of a more inclusive node, "n1" (Fig. 2g). 316

An Example with the Family of True Finches

Creating a query. To obtain ages for all species within the family of true finches, 318 Fringillidae, we ran a DateLife query using the "get species from taxon" flag, which gets all 319 recognized species names within a named group from a taxonomy of choice. Following the 320 NCBI taxonomy, our DateLife query has 289 Fringillidae species names. This 321 taxon-constrained approach implies that the full DateLife analysis will be performed using a 322 tree topology and ages available for species names from a given taxonomic group, which do 323 not necessarily correspond to a monophyletic group. Users can change this behavior by 324 providing all species names corresponding to a monophyletic group as input for a DateLife 325 search, or a monophyletic tree to construct a DateLife summary. 326

Searching the database. Next, we used the processed species names in our 327 DateLife query to identify chronograms with at least two Fringillidae species as tip taxa. 328 The DateLife search identified 19 chronograms matching this criteria, published in 13 320 different studies (Barker et al., 2013, 2015; Burns et al., 2014; Claramunt & Cracraft, 2015; 330 Gibb et al., 2015; Hedges et al., 2015; Hooper & Price, 2017; Jetz et al., 2012; Kimball et al., 331 2019; Oliveros et al., 2019; Price et al., 2014; Roquet, Lavergne, & Thuiller, 2014; Uyeda, 332 Pennell, Miller, Maia, & McClain, 2017). Once identified, DateLife pruned these matching 333 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 (Supplementary 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 (Supplementary Fig. S2; Open Tree Of Life et al., 2019).

All 19 source chronograms (Supplementary Figs. S5-S23) were congruified to 346 OpenTree's topology shown in Supplementary Figure S2, reducing the original 1,206 node 347 age data set to 818 different data points (Supplementary Table S2) that could be used as 348 calibrations for that chosen topology. The congruent node age data points were summarized 349 for each node, resulting in 194 summary node ages. From these 21 were excluded as 350 secondary calibrations because they were older than the ancestral node. The remaining 173 351 summary node ages were used as secondary calibrations to obtain a fully dated (and 352 resolved) phylogeny with the program BLADJ (Supplementary Figure S3). Results for a 353 subgroup are shown in Figure 3. 354

CROSS-VALIDATION TEST

We performed a cross validation test of a DateLife analysis using the Fringillidae 356 source chronograms obtained above (Supplementary Figs. S5-S23). As inputs for a DateLife 357 analysis, we used all individual tree topologies from each of the 19 source chronograms from 358 13 studies, treating their node ages as unknown. We congruified node ages extracted from 359 chronograms from all other studies upon the individual topologies, effectively excluding 360 original ages from each topology. Finally, average node ages per node were applied as 361 secondary calibrations and smoothed with the BLADJ algorithm. We found that node ages 362 from the original studies, and ages estimated using all other age data available are generally correlated (Fig. 4). For five studies, DateLife tended to underestimate ages for topologically deeper nodes (those with many descendant taxa, aka "closer to the root") relative to the original estimate, and overestimate ages for nodes closer to the tips. Accordingly, root ages 366 are generally older in the original study than estimated using cross-validated ages 367 (Supplementary Fig. S4). In general, topologically deeper nodes display the largest age

variation between node ages from the original chronograms and ages summarized with

DateLife.

DISCUSSION

372

373

374

375

376

377

DateLife's goal is to improve availability, accessibility, and reusability 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, use 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 that 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. 378 Incorporation of more chronograms into the database will continue to improve DateLife's 379 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 381 designed and implemented. However, Dryad's metadata system has no information to 382 automatically detect branch length units, and those would still need to be determined 383 manually by a human curator. We would like to emphasize on the importance of sharing 384 chronogram data, including systematically curated metadata, into open repositories, such as 385 OpenTree's Phylesystem (McTavish et al., 2015) for the benefit of research and the scientific 386 community as a whole. Another important source of expert data on time of lineage 387 divergence is TimeTree's database (Hedges, Dudley, & Kumar, 2006), which holds 388 chronograms from more than 4k published studies, and is fully browsable using its graphical user interface (timetree.org). TimeTree's chronogram database was not accessible in computer readable format until very recently (Kumar et al., 2022), when its terms of use and 391 website application were updated, now allowing some kinds of reuse, but not redistribution. 392 The inaccessibility of TimeTree's database was an inspiration for the DateLife project, 393 which was born as a prototype tool initially developed over a series of hackathons at the

National Evolutionary Synthesis Center, NC, USA (Stoltzfus et al., 2013), as the need to make scientific information that is funded by the public practically available to the public was acknowledged and prioritized.

As we envision that DateLife will have many interesting applications in research and 398 beyond, we emphasize that DateLife's results –as well as any insights gleaned from them, 390 largely depend on the quality of the source chronograms: low quality chronograms will 400 produce low quality results. The "garbage in, garbage out" problem has long been recognised 401 in supertree methods for summarizing phylogenetic trees (Bininda-Emonds et al., 2004). We 402 note that this is a surfacing issue of any automated tool for biological data analysis. For 403 example, DNA riddled with sequencing errors will produce generally poor alignments that will return biased evolutionary hypothesis, independently of the quality of the analysis 405 software used. Again, we urge readers and DateLife users to explore all input chronograms 406 before using a summary chronogram resulting from a DateLife workflow. 407

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 the best 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,
for example, the robust ongoing debate about crown group age of angiosperms
(Barba-Montoya, Reis, Schneider, Donoghue, & Yang, 2018; Magallón, Gómez-Acevedo,
Sánchez-Reyes, & Hernández-Hernández, 2015; Ramshaw et al., 1972; Sanderson & Doyle,

2001; Sauquet, Ramírez-Barahona, & Magallón, 2021). Alternative source chronograms available for the same taxa have potentially been estimated implementing different types of 421 calibrations, which affects the resulting node age estimates. For example, in the DateLife 422 analysis of the Fringillidae shown above, the chronograms from one study (Burns et al., 2014) 423 were inferred using molecular substitution rate estimates across birds (Weir & Schluter, 424 2008), and have much older age estimates for the same nodes than chronograms that were 425 inferred using fossil calibrations (Figs. 3, 4c, Supplementary Figs. S4c, S10). Another source 426 of conflict in estimated node ages can arise from different placements for the same calibration, 427 which would imply fundamentally distinct evolutionary hypotheses (Antonelli et al., 2017). 428 For example, two independent researchers working on the same clade should both carefully 429 select and justify their choices of fossil calibration placement. Yet, if one researcher concludes 430 that a fossil should calibrate the ingroup of a clade, while another researcher concludes that 431 the same fossil should calibrate the outgroup of the clade, the resulting age estimates will 432 differ, as the placement of calibrations as stem or crown group is known to significantly affect estimates of time of lineage divergence (Sauquet, 2013). Finally, placement of calibrations also affects uncertainty of node age estimates. For example, nodes that are sandwiched 435 between a calibrated node and a calibrated root have less freedom of movement and hence narrower confidence intervals (Vos & Mooers, 2004), which inflates precision for nodes 437 without calibrations but does not 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.

DateLife constructs summary chronograms using node ages extracted from existing 446 chronograms, i.e. secondary calibrations. In general, the scientific community has more 447 confidence in chronograms using primary calibrations, where the dated tree is generated from 448 a single analysis where carefully chosen fossil calibrations are the source of absolute time 449 information, than in analyses dated using secondary calibrations (Antonelli et al., 2017; 450 Garzón-Orduña, Silva-Brandão, Willmott, Freitas, & Brower, 2015; Graur & Martin, 2004; 451 Sauguet, 2013; Sauguet et al., 2012; Schenk, 2016; Shaul & Graur, 2002). However, 452 implementation of primary calibrations is difficult: it requires specialized expertise and 453 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 example done by Antonelli et al. (2017). The Fossil Calibration Database provides data for 220 primary calibration points encompassing flowering plants and metazoans, that have been 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 appropriate nodes in their phylogenies which correspond with fossils available in the 463 database. 464

Recently, Powell, Waskin, and Battistuzzi (2020) showed in a simulation study that secondary calibrations using node ages based on previous molecular clock analyses can be as good as primary calibrations. Using several secondary calibrations (as opposed to just one) can provide sufficient information to alleviate or even neutralize potential biases (Graur & Martin, 2004; Sauquet, 2013; Shaul & Graur, 2002). Our cross validation analysis 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. S5-S23). We also note that DateLife estimates for nodes closer to the root tend to be slightly

younger than ages from the original studies. In contrast, nodes closer to the tips tend to be 473 slightly older when estimated using our secondary calibrations than ages from the original 474 studies. The only exception to this trend was observed in Burns et al. (2014) chronogram, 475 which generally displays much younger node ages when estimated using secondary 476 calibrations than the original study (Supplementary Figs. S4c, S10), supporting previous 477 observations (Sauquet et al., 2012; Schenk, 2016). However, these younger dates are more 478 likely an example of how multiple secondary calibrations can correct erroneous estimates, as 470 dates on the Burns et al. (2014) tree were obtained using a single secondary calibration 480 based on a previously estimated molecular evolution rate across birds from Weir and 481 Schluter (2008), and appear as major outliers compared to alternate estimates for the same 482 nodes based on primary fossil calibrations (Fig. 3, Supplementary Fig. S3). 483

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.

500

501

502

503

514

523

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 is a common 504 resource in research, as it provides the opportunity to work with a chronogram that reflects a 505 unified evolutionary history for a lineage, by putting together evidence from different 506 hypotheses. For example, the largest, and taxonomically broadest chronogram currently 507 available from OpenTree was constructed summarizing age data from 2,274 published 508 chronograms using NCBI's taxonomic tree as backbone (Hedges et al., 2015), which has been 500 widely reused for research. Finally, we note that summarizing chronograms should be done 510 with caution, as it may amplify the effect of uncertainty and errors in source data, and blur 511 parts of the evolutionary history of a lineage that might only be reflected in source chronograms and lost on the summary chronogram (Sauquet et al., 2021). 513

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 525 was simulated; Rabosky et al. (2018) created a chronogram of 31,536 ray-finned fishes, of 526 which only 37% had molecular data; Smith and Brown (2018) constructed a chronogram of 527 353,185 seed plants where only 23% had molecular data. Stochastically resolved chronograms 528 can return diversification rates estimates that appear less biased than those estimated from 529 their incompletely sampled counterparts, even with methods that account for missing 530 lineages by using sampling fractions (Chang et al., 2020; Cusimano et al., 2012), but can also 531 introduce spurious patterns of early bursts of diversification (Cusimano & Renner, 2010; Sun 532 et al., 2020). 533

Taxonomy-based stochastic polytomy resolvers also introduce topological differences in 534 phylogenetic trees. The study of macroevolutionary processes largely depends on an 535 understanding of the timing of species diversification events, and different phylogenetic and 536 chronogram hypothesis can provide very different overviews of the macroevolutionary history 537 of a biological group. For example, alternative topologies in chronograms from the same 538 biological group can infer very different species diversification patterns (Rabosky, 2015; Title 539 & Rabosky, 2016). Similarly, there are worries that patterns of morphological evolution 540 cannot be accurately inferred with phylogenies that have been resolved stochastically over a 541 taxonomic backbone, as any patterns would be erased by randomization (Rabosky, 2015). 542 We note that the same applies for geography- and morphology-dependent diversification 543 analysis. Hence, we suggest that phylogenies that have been processed with taxonomy-based stochastic polytomy resolvers, including certain summary chronograms from a DateLife 545 analysis, can be useful as null or neutral models, representing the case of a diversification process that is independent of traits and geographical scenario.

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
consistency of DateLife's results with prior work.

CONCLUSIONS CONCLUSIONS

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 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.

The DateLife project allows for easy and fast summarization of public and 565 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 567 to get an informed picture of the state of knowledge for the time frame of evolution of 568 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 572 variation in expert estimations of time of divergence, and foster exploration of the effect of 573 alternative divergence time hypotheses on the results of analyses, nurturing a culture of more 574 cautious interpretation of evolutionary results. 575

Availability

576

587

The DateLife software is free and open source. It can be used online through its R 577 shiny web application hosted at http://www.datelife.org, and locally through the datelife 578 R package, available from Zenodo (https://doi.org/10.5281/zenodo.593938 and the CRAN 579 repository (Sanchez-Reyes et al., 2022). DateLife's web application is maintained using 580 RStudio's shiny server and the shiny package open infrastructure, as well as Docker and 581 OpenTree's infrastructure (datelife.opentreeoflife.org). datelife's stable version can be 582 installed from the CRAN repository using the command install.packages(pkgs = 583 "datelife") from within R. Development versions are available from DateLife's GitHub 584 repository (https://github.com/phylotastic/datelife) and can be installed using the 585 command devtools::install github("phylotastic/datelife"). 586

SUPPLEMENTARY MATERIAL

Supplementary material, including figures, tables, code, biological examples, 588 benchmark results, data files and online-only appendices, can be viewed and downloaded 589 from the Dryad data repository (https://doi.org/10.5061/dryad.cnp5hqc6w), as well as from 590 the Zenodo stable repositories that host the reproducible manuscript 591 (https://doi.org/10.5281/zenodo.7435094), the biological examples 592 (https://doi.org/10.5281/zenodo.7435101), the software benchmark (https://doi.org/10.5281/zenodo.7435106), and the figures (https://doi.org/10.5281/zenodo.6683667). Dryad's data publication fee is covered by the Society of Systematic Biologists. Development versions corresponding to all of the above are hosted on GitHub, accessible at https://github.com/LunaSare/datelifeMS1, 597 https://github.com/LunaSare/datelife examples, and 598 https://github.com/LunaSare/datelife benchmark. 590

600 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 reviewer, 606 for comments that greatly improved this manuscript. We thank colleagues from the O'Meara 607 Lab at the University of Tennessee Knoxville for suggestions, discussions and software 608 testing. The late National Evolutionary Synthesis Center (NESCent), which sponsored 609 hackathons that led to initial work on this project. The team that assembled DateLife's first 610 proof of concept: Tracy Heath, Jonathan Eastman, Peter Midford, Joseph Brown, Matt Pennell, Mike Alfaro, and Luke Harmon. The Open Tree of Life project that provides the open, metadata rich repository of trees used to construct DateLife's chronogram database. 613 The many scientists who publish their chronograms in an open, reusable form, and the 614 scientists who curate them for deposition in the Open Tree of Life repository. The NSF for 615 funding nearly all the above, in addition to the ABI grant that funded this project itself.

References

- Agnarsson, I., & Miller, J. A. (2008). Is ACCTRAN better than DELTRAN? *Cladistics*, 24 (6), 1032–1038.
- Alström, P., Hooper, D. M., Liu, Y., Olsson, U., Mohan, D., Gelang, M., ... Price, T. D.
- (2014). Discovery of a relict lineage and monotypic family of passerine birds. Biology Letters, 10(3), 20131067.
- Ané, C., Eulenstein, O., Piaggio-Talice, R., & Sanderson, M. J. (2009). Groves of phylogenetic trees. *Annals of Combinatorics*, 13(2), 139–167.
- Antonelli, A., Hettling, H., Condamine, F. L., Vos, K., Nilsson, R. H., Sanderson, M. J., ...
- Vos, R. A. (2017). Toward a self-updating platform for estimating rates of speciation
- and migration, ages, and relationships of Taxa. Systematic Biology, 66(2), 153–166.
- 628 https://doi.org/10.1093/sysbio/syw066
- Archie, J., Day, W. H., Felsenstein, J., Maddison, W., Meacham, C., Rohlf, F. J., &
- Swofford, D. (1986). The Newick tree format. Retrieved from
- {https://evolution.genetics.washington.edu/phylip/newicktree.html}
- Avibase. (2022). Yellow-throated Bunting. Avibase the World Bird Database, (Online
- Resource). Retrieved from
- 634 {https://avibase.bsc-eoc.org/species.jsp?lang=EN&avibaseid=82D1EE0049D8D927}
- Bapst, D. W. (2012). Paleotree: An R package for paleontological and phylogenetic analyses
- of evolution. Methods in Ecology and Evolution, 3(5), 803–807.
- 637 https://doi.org/10.1111/j.2041-210X.2012.00223.x
- Barba-Montoya, J., Reis, M. dos, Schneider, H., Donoghue, P. C., & Yang, Z. (2018).
- Constraining uncertainty in the timescale of angiosperm evolution and the veracity of

- a cretaceous terrestrial revolution. New Phytologist, 218(2), 819–834.
- Barker, F. K. (2014). Mitogenomic data resolve basal relationships among passeriform and passeridan birds. *Molecular Phylogenetics and Evolution*, 79, 313–324.
- Barker, F. K., Burns, K. J., Klicka, J., Lanyon, S. M., & Lovette, I. J. (2013). Going to
 extremes: Contrasting rates of diversification in a recent radiation of new world
 passerine birds. Systematic Biology, 62(2), 298–320.
- Barker, F. K., Burns, K. J., Klicka, J., Lanyon, S. M., & Lovette, I. J. (2015). New insights
 into new world biogeography: An integrated view from the phylogeny of blackbirds,
 cardinals, sparrows, tanagers, warblers, and allies. The Auk: Ornithological Advances,
 132(2), 333–348.
- Barker, F. K., Cibois, A., Schikler, P., Feinstein, J., & Cracraft, J. (2004). Phylogeny and diversification of the largest avian radiation. *Proceedings of the National Academy of Sciences*, 101(30), 11040–11045.
- Beresford, P., Barker, F., Ryan, P., & Crowe, T. (2005). African endemics span the tree of songbirds (passeri): Molecular systematics of several evolutionary "enigmas".

 Proceedings of the Royal Society B: Biological Sciences, 272(1565), 849–858.
- Bininda-Emonds, O. R., Jones, K. E., Price, S. A., Cardillo, M., Grenyer, R., & Purvis, A.
 (2004). Garbage in, garbage out: Data issues in supertree construction. *Phylogenetic Supertrees: Combining Information to Reveal the Tree of Life*, 267–280.
- Bodenhofer, U., Bonatesta, E., Horejš-Kainrath, C., & Hochreiter, S. (2015). Msa: An r package for multiple sequence alignment. *Bioinformatics*, 31(24), 3997–3999.
- Boyle, B., Hopkins, N., Lu, Z., Raygoza Garay, J. A., Mozzherin, D., Rees, T., ... Enquist,
 B. J. (2013). The taxonomic name resolution service: An online tool for automated

- standardization of plant names. BMC Bioinformatics, 14(1).
- https://doi.org/10.1186/1471-2105-14-16
- Britton, T., Anderson, C. L., Jacquet, D., Lundqvist, S., & Bremer, K. (2007). Estimating
- Divergence Times in Large Phylogenetic Trees. Systematic Biology, 56 (788777878),
- 667 741–752. https://doi.org/10.1080/10635150701613783
- Bryson Jr, R. W., Chaves, J., Smith, B. T., Miller, M. J., Winker, K., Pérez-Emán, J. L., &
- Klicka, J. (2014). Diversification across the new world within the 'blue' cardinalids
- (aves: Cardinalidae). Journal of Biogeography, 41(3), 587–599.
- Burleigh, J. G., Kimball, R. T., & Braun, E. L. (2015). Building the avian tree of life using a
- large-scale, sparse supermatrix. Molecular Phylogenetics and Evolution, 84, 53–63.
- Burns, K. J., Shultz, A. J., Title, P. O., Mason, N. A., Barker, F. K., Klicka, J., ... Lovette,
- I. J. (2014). Phylogenetics and diversification of tanagers (passeriformes:
- Thraupidae), the largest radiation of neotropical songbirds. Molecular Phylogenetics
- and Evolution, 75, 41–77.
- ⁶⁷⁷ Chamberlain, S. (2018). bold: Interface to Bold Systems API. Retrieved from
- 678 https://CRAN.R-project.org/package=bold
- ⁶⁷⁹ Chamberlain, S. A., & Szöcs, E. (2013). taxize: taxonomic search and retrieval in R [version
- 2; referees: 3 approved]. F1000Research, 2(191), 1–29.
- https://doi.org/10.12688/f1000research.2-191.v2
- 682 Chang, J., Rabosky, D. L., & Alfaro, M. E. (2020). Estimating diversification rates on
- incompletely sampled phylogenies: Theoretical concerns and practical solutions.
- Systematic Biology, 69(3), 602-611.
- ⁶⁸⁵ Chaves, J. A., Hidalgo, J. R., & Klicka, J. (2013). Biogeography and evolutionary history of

- the n eotropical genus s altator (a ves: T hraupini). Journal of Biogeography, 40(11), 2180–2190.
- Claramunt, S., & Cracraft, J. (2015). A new time tree reveals earth history's imprint on the evolution of modern birds. *Science Advances*, 1(11), e1501005.
- Criscuolo, A., Berry, V., Douzery, E. J., & Gascuel, O. (2006). SDM: A fast distance-based approach for (super)tree building in phylogenomics. Systematic Biology, 55(5),

 740–755. https://doi.org/10.1080/10635150600969872
- Cusimano, N., & Renner, S. S. (2010). Slowdowns in diversification rates from real phylogenies may not be real. Systematic Biology, 59(4), 458–464.
- Cusimano, N., Stadler, T., & Renner, S. S. (2012). A new method for handling missing
 species in diversification analysis applicable to randomly or nonrandomly sampled
 phylogenies. Systematic Biology, 61(5), 785–792.
- Delsuc, F., Philippe, H., Tsagkogeorga, G., Simion, P., Tilak, M.-K., Turon, X., . . . Douzery,
 E. J. (2018). A phylogenomic framework and timescale for comparative studies of
 tunicates. *BMC Biology*, 16(1), 1–14.
- Eastman, J. M., Harmon, L. J., & Tank, D. C. (2013). Congruification: Support for time scaling large phylogenetic trees. *Methods in Ecology and Evolution*, 4(7), 688–691. https://doi.org/10.1111/2041-210X.12051
- Edgar, R. C. (2004). MUSCLE: Multiple sequence alignment with high accuracy and high throughput. *Nucleic Acids Research*, 32(5), 1792–1797.
- Felsenstein, J. (1985). Phylogenies and the Comparative Method. *The American Naturalist*,

 125(1), 1–15. Retrieved from http://www.jstor.org/stable/2461605
- ⁷⁰⁸ Forest, F., Savolainen, V., Chase, M. W., Lupia, R., Bruneau, A., & Crane, P. R. (2005).

- Teasing apart molecular-versus fossil-based error estimates when dating phylogenetic trees: A case study in the birch family (betulaceae). Systematic Botany, 30(1), 118–133.
- Freckleton, R. P., Harvey, P. H., & Pagel, M. (2002). Phylogenetic analysis and comparative data: A test and review of evidence. *The American Naturalist*.
- Garzón-Orduña, I. J., Silva-Brandão, K. L., Willmott, K. R., Freitas, A. V., & Brower, A. V. (2015). Incompatible ages for clearwing butterflies based on alternative secondary calibrations. Systematic Biology, 64(5), 752–767.
- GBIF Secretariat. (2022). GBIF Backbone Taxonomy. *Checklist dataset*, (Online Resource accessed via GBIF.org). Retrieved from {https://doi.org/10.15468/39omei }
- Gibb, G. C., England, R., Hartig, G., McLenachan, P. A., Taylor Smith, B. L., McComish,
 B. J., ... Penny, D. (2015). New zealand passerines help clarify the diversification of
 major songbird lineages during the oligocene. Genome Biology and Evolution, 7(11),
 2983–2995.
- Graur, D., & Martin, W. (2004). Reading the entrails of chickens: Molecular timescales of evolution and the illusion of precision. *TRENDS in Genetics*, 20(2), 80–86.
- Hackett, S. J., Kimball, R. T., Reddy, S., Bowie, R. C., Braun, E. L., Braun, M. J., ...
 others. (2008). A phylogenomic study of birds reveals their evolutionary history.

 Science, 320(5884), 1763–1768.
- Harvey, P. H., Pagel, M. D., & others. (1991). The comparative method in evolutionary
 biology (Vol. 239). Oxford university press Oxford.
- Hedges, S. B., Dudley, J., & Kumar, S. (2006). TimeTree: A public knowledge-base of divergence times among organisms. *Bioinformatics*, 22(23), 2971–2972.

- https://doi.org/10.1093/bioinformatics/btl505
- Hedges, S. B., Marin, J., Suleski, M., Paymer, M., & Kumar, S. (2015). Tree of life reveals clock-like speciation and diversification. *Molecular Biology and Evolution*, 32(4),
- 735 835-845. https://doi.org/10.1093/molbev/msv037
- Heibl, C. (2008). PHYLOCH: R language tree plotting tools and interfaces to diverse

 phylogenetic software packages. Retrieved from
- http://www.christophheibl.de/Rpackages.html
- Hipsley, C. A., & Müller, J. (2014). Beyond fossil calibrations: Realities of molecular clock practices in evolutionary biology. *Frontiers in Genetics*, 5, 138.
- Hooper, D. M., & Price, T. D. (2017). Chromosomal inversion differences correlate with range overlap in passerine birds. *Nature Ecology & Evolution*, 1(10), 1526.
- Huelsenbeck, J. P., & Ronquist, F. (2001). MRBAYES: Bayesian inference of phylogenetic trees. *Bioinformatics*, 17(8), 754–755.
- https://doi.org/10.1093/bioinformatics/17.8.754
- Jetz, W., Thomas, G., Joy, J. J., Hartmann, K., & Mooers, A. (2012). The global diversity of birds in space and time. *Nature*, 491 (7424), 444–448.
- https://doi.org/10.1038/nature11631
- Johansson, U. S., Fjeldså, J., & Bowie, R. C. (2008). Phylogenetic relationships within

 passerida (aves: Passeriformes): A review and a new molecular phylogeny based on

 three nuclear intron markers. *Molecular Phylogenetics and Evolution*, 48(3), 858–876.
- Katoh, K., Asimenos, G., & Toh, H. (2009). Multiple alignment of dna sequences with mafft.

 In *Bioinformatics for dna sequence analysis* (pp. 39–64). Springer.
- Kimball, R. T., Oliveros, C. H., Wang, N., White, N. D., Barker, F. K., Field, D. J., ...

- others. (2019). A phylogenomic supertree of birds. Diversity, 11(7), 109.
- Klicka, J., Barker, F. K., Burns, K. J., Lanyon, S. M., Lovette, I. J., Chaves, J. A., & Bryson
- Jr, R. W. (2014). A comprehensive multilocus assessment of sparrow (aves:
- Passerellidae) relationships. Molecular Phylogenetics and Evolution, 77, 177–182.
- Ksepka, D. T., Benton, M. J., Carrano, M. T., Gandolfo, M. A., Head, J. J., Hermsen, E. J.,
- ... others. (2011). Synthesizing and databasing fossil calibrations: Divergence dating
- and beyond. The Royal Society.
- Ksepka, D. T., Parham, J. F., Allman, J. F., Benton, M. J., Carrano, M. T., Cranston, K.
- A., ... others. (2015). The fossil calibration database—a new resource for divergence
- dating. Systematic Biology, 64(5), 853-859.
- Kumar, S., Suleski, M., Craig, J. M., Kasprowicz, A. E., Sanderford, M., Li, M., ... Hedges,
- S. B. (2022). TimeTree 5: An expanded resource for species divergence times.
- Molecular Biology and Evolution, 39(8), msac174.
- Lamichhaney, S., Berglund, J., Almén, M. S., Maqbool, K., Grabherr, M., Martinez-Barrio,
- A., ... others. (2015). Evolution of darwin's finches and their beaks revealed by
- genome sequencing. *Nature*, 518(7539), 371–375.
- Laubichler, M. D., & Maienschein, J. (2009). Form and function in developmental evolution.
- Cambridge University Press.
- 773 Lepage, D. (2004). Avibase: The world bird database. Bird Studies Canada.
- Lepage, D., Vaidya, G., & Guralnick, R. (2014). Avibase—a database system for managing
- and organizing taxonomic concepts. ZooKeys, (420), 117.
- Lerner, H. R., Meyer, M., James, H. F., Hofreiter, M., & Fleischer, R. C. (2011). Multilocus
- resolution of phylogeny and timescale in the extant adaptive radiation of hawaiian

- noneycreepers. Current Biology, 21 (21), 1838–1844.
- Lovette, I. J., Pérez-Emán, J. L., Sullivan, J. P., Banks, R. C., Fiorentino, I.,
- Córdoba-Córdoba, S., ... others. (2010). A comprehensive multilocus phylogeny for
- the wood-warblers and a revised classification of the parulidae (aves). *Molecular*
- Phylogenetics and Evolution, 57(2), 753-770.
- Magallon, S., & Sanderson, M. (2001). Absolute diversification rates in angiosperm clades.
- Evolution, 55(9), 1762–1780.
- Magallón, S. (2010). Using fossils to break long branches in molecular dating: A comparison
- of relaxed clocks applied to the origin of angiosperms. Systematic Biology, 59(4),
- ₇₈₇ 384–399.
- Magallón, S., Gómez-Acevedo, S., Sánchez-Reyes, L. L., & Hernández-Hernández, T. (2015).
- A metacalibrated time-tree documents the early rise of flowering plant phylogenetic
- diversity. New Phytologist, 207(2), 437-453.
- McTavish, E. J., Hinchliff, C. E., Allman, J. F., Brown, J. W., Cranston, K. A., Holder, M.
- T., ... Smith, S. (2015). Phylesystem: A git-based data store for community-curated
- phylogenetic estimates. *Bioinformatics*, 31 (17), 2794–2800.
- Michonneau, F., Brown, J. W., & Winter, D. J. (2016). rotl: an R package to interact with
- the Open Tree of Life data. Methods in Ecology and Evolution, 7(12), 1476–1481.
- 796 https://doi.org/10.1111/2041-210X.12593
- Morlon, H. (2014). Phylogenetic approaches for studying diversification. Ecology Letters,
- 798 17(4), 508–525. https://doi.org/10.1111/ele.12251
- Moyle, R. G., Oliveros, C. H., Andersen, M. J., Hosner, P. A., Benz, B. W., Manthey, J. D.,
- 800 ... Faircloth, B. C. (2016). Tectonic collision and uplift of wallacea triggered the

- global songbird radiation. Nature Communications, 7(1), 1–7.
- Oliveros, C. H., Field, D. J., Ksepka, D. T., Barker, F. K., Aleixo, A., Andersen, M. J., ...
- others. (2019). Earth history and the passerine superradiation. Proceedings of the
- National Academy of Sciences, 116(16), 7916-7925.
- Ooms, J., & Chamberlain, S. (2018). *Phylocomr: Interface to 'phylocom'*. Retrieved from https://CRAN.R-project.org/package=phylocomr
- Open Tree Of Life, Redelings, B., Cranston, K. A., Allman, J., Holder, M. T., & McTavish,
- E. J. (2016). Open Tree of Life APIs v3.0. Open Tree of Life Project, (Online
- Resources). Retrieved from
- 810 {https://github.com/OpenTreeOfLife/germinator/wiki/Open-Tree-of-Life-Web-
- 811 APIs}
- Open Tree Of Life, Redelings, B., Sánchez Reyes, L. L., Cranston, K. A., Allman, J., Holder,
- M. T., & McTavish, E. J. (2019). Open tree of life synthetic tree v12.3. Zenodo.
- Retrieved from https://doi.org/10.5281/zenodo.3937742
- Ödeen, A., Håstad, O., & Alström, P. (2011). Evolution of ultraviolet vision in the largest avian radiation-the passerines. *BMC Evolutionary Biology*, 11(1), 1–8.
- Paradis, E., Claude, J., & Strimmer, K. (2004). APE: analyses of phylogenetics and evolution in R language. *Bioinformatics*, 20(2), 289–290.
- Parchman, T. L., Benkman, C. W., & Mezquida, E. T. (2007). Coevolution between hispaniolan crossbills and pine: Does more time allow for greater phenotypic escalation at lower latitude? *Evolution*, 61(9), 2142–2153.
- Päckert, M., Martens, J., Sun, Y.-H., Severinghaus, L. L., Nazarenko, A. A., Ting, J., ...
- Tietze, D. T. (2012). Horizontal and elevational phylogeographic patterns of

- himalayan and southeast asian forest passerines (aves: Passeriformes). Journal of
 Biogeography, 39(3), 556–573.
- Pennell, M. W., Eastman, J. M., Slater, G. J., Brown, J. W., Uyeda, J. C., FitzJohn, R. G.,

 Harmon, L. J. (2014). Geiger v2. 0: An expanded suite of methods for fitting

 macroevolutionary models to phylogenetic trees. *Bioinformatics*, 30(15), 2216–2218.
- Posadas, P., Crisci, J. V., & Katinas, L. (2006). Historical biogeography: A review of its basic concepts and critical issues. *Journal of Arid Environments*, 66(3), 389–403.
- Powell, A. F., Barker, F. K., Lanyon, S. M., Burns, K. J., Klicka, J., & Lovette, I. J. (2014).

 A comprehensive species-level molecular phylogeny of the new world blackbirds

 (icteridae). *Molecular Phylogenetics and Evolution*, 71, 94–112.
- Powell, C. L. E., Waskin, S., & Battistuzzi, F. U. (2020). Quantifying the error of secondary vs. Distant primary calibrations in a simulated environment. *Frontiers in Genetics*, 11, 252.
- Price, T. D., Hooper, D. M., Buchanan, C. D., Johansson, U. S., Tietze, D. T., Alström, P.,
 others. (2014). Niche filling slows the diversification of himalayan songbirds.

 Nature, 509 (7499), 222.
- Pulgarín-R, P. C., Smith, B. T., Bryson Jr, R. W., Spellman, G. M., & Klicka, J. (2013).
 Multilocus phylogeny and biogeography of the new world pheucticus grosbeaks (aves:
 Cardinalidae). Molecular Phylogenetics and Evolution, 69(3), 1222–1227.
- Rabosky, D. L. (2015). No substitute for real data: A cautionary note on the use of
 phylogenies from birth-death polytomy resolvers for downstream comparative
 analyses. *Evolution*, 69(12), 3207–3216.
- Rabosky, D. L., Chang, J., Title, P. O., Cowman, P. F., Sallan, L., Friedman, M., ... others.

- (2018). An inverse latitudinal gradient in speciation rate for marine fishes. *Nature*, 559 (7714), 392.
- Ramshaw, J., Richardson, D., Meatyard, B., Brown, R., Richardson, M., Thompson, E., & Boulter, D. (1972). The time of origin of the flowering plants determined by using amino acid sequence data of cytochrome c. New Phytologist, 71(5), 773–779.
- Ratnasingham, S., & Hebert, P. D. (2007). BOLD: The barcode of life data system

 (http://www. Barcodinglife. Org). Molecular Ecology Notes, 7(3), 355–364.
- R Core Team. (2018). R: a language and environment for statistical computing. Vienna,

 Austria: R Foundation for Statistical Computing.
- Rees, & Cranston, K. (2017). Automated assembly of a reference taxonomy for phylogenetic data synthesis. *Biodiversity Data Journal*, (5).
- Rees, Vandepitte, L., Decock, W., & Vanhoorne, B. (2017). IRMNG 2006–2016: 10 Years of a Global Taxonomic Database. *Biodiversity Informatics*, 12.
- Revell, L. J. (2012). Phytools: An r package for phylogenetic comparative biology (and other things). Methods in Ecology and Evolution, 3, 217–223.
- Ronquist, F., & Huelsenbeck, J. P. (2003). MrBayes 3: Bayesian phylogenetic inference under mixed models. *Bioinformatics*, 19(12), 1572–1574. https://doi.org/10.1093/bioinformatics/btg180
- Roquet, C., Lavergne, S., & Thuiller, W. (2014). One tree to link them all: A phylogenetic dataset for the european tetrapoda. *PLoS Currents*, 6.
- Sanchez-Reyes, L. L., & O'Meara, B. (2022). datelifeplot: Methods to plot chronograms
 and outputs of the datelife package. R Package Release V0.2.2. Retrieved from
 https://zenodo.org/badge/latestdoi/381501451

- Sanchez-Reyes, L. L., O'Meara, B., Eastman, J., Heath, T., Wright, A., Schliep, K., ...
- Alfaro, M. (2022). datelife: Scientific Data on Time of Lineage Divergence for Your
- Taxa. In R package version 0.6.6. Retrieved from
- https://CRAN.R-project.org/package=datelife and
- https://doi.org/10.5281/zenodo.593938
- Sanderson, M. (2002). Estimating Absolute Rates of Molecular Evolution and Divergence
- Times: A Penalized Likelihood Approach. Molecular Biology and Evolution, 19(1),
- 877 101–109. https://doi.org/10.1093/oxfordjournals.molbev.a003974
- Sanderson, M., & Doyle, J. (2001). Sources of error and confidence intervals in estimating
- the age of angiosperms from rbcL and 18S rDNA data. American Journal of Botany,
- 880 88(8), 1499–1516.
- Sauquet, H. (2013). A practical guide to molecular dating. Comptes Rendus Palevol, 12(6),
- 882 355–367.
- Sauquet, H., Ho, S. Y. W., Gandolfo, M. a, Jordan, G. J., Wilf, P., Cantrill, D. J., ...
- Udovicic, F. (2012). Testing the impact of calibration on molecular divergence times
- using a fossil-rich group: the case of Nothofagus (Fagales). Systematic Biology, 61(2),
- 289–313. https://doi.org/10.1093/sysbio/syr116
- Sauquet, H., Ramírez-Barahona, S., & Magallón, S. (2021). The age of flowering plants is
- unknown.
- Schenk, J. J. (2016). Consequences of secondary calibrations on divergence time estimates.
- 890 PLoS ONE, 11(1). https://doi.org/10.1371/journal.pone.0148228
- Schliep, K. P. (2011). Phangorn: Phylogenetic analysis in r. Bioinformatics, 27(4), 592–593.
- Schoch, C. L., Ciufo, S., Domrachev, M., Hotton, C. L., Kannan, S., Khovanskaya, R., . . .

- others. (2020). NCBI Taxonomy: a Comprehensive Update on Curation, Resources and Tools. *Database*, 2020.
- Selvatti, A. P., Gonzaga, L. P., & Moraes Russo, C. A. de. (2015). A paleogene origin for crown passerines and the diversification of the oscines in the new world. *Molecular Phylogenetics and Evolution*, 88, 1–15.
- Shaul, S., & Graur, D. (2002). Playing chicken (gallus gallus): Methodological inconsistencies of molecular divergence date estimates due to secondary calibration points. *Gene*, 300 (1-2), 59–61.
- Smith, S., & Brown, J. (2018). Constructing a broadly inclusive seed plant phylogeny.

 American Journal of Botany, 105(3), 302–314.
- Smith, S., & O'Meara, B. (2012). TreePL: Divergence time estimation using penalized likelihood for large phylogenies. *Bioinformatics*, 28(20), 2689–2690. https://doi.org/10.1093/bioinformatics/bts492
- Stoltzfus, A., Lapp, H., Matasci, N., Deus, H., Sidlauskas, B., Zmasek, C. M., ... Jordan, G. (2013). Phylotastic! Making tree-of-life knowledge accessible, reusable and convenient.

 BMC Bioinformatics, 14. https://doi.org/10.1186/1471-2105-14-158
- Sun, M., Folk, R. A., Gitzendanner, M. A., Soltis, P. S., Chen, Z., Soltis, D. E., & Guralnick,
 R. P. (2020). Estimating rates and patterns of diversification with incomplete
 sampling: A case study in the rosids. American Journal of Botany, 107(6), 895–909.
- Tietze, D. T., Päckert, M., Martens, J., Lehmann, H., & Sun, Y.-H. (2013). Complete
 phylogeny and historical biogeography of true rosefinches (aves: Carpodacus).

 Zoological Journal of the Linnean Society, 169(1), 215–234.
- Title, P. O., & Rabosky, D. L. (2016). Do Macrophylogenies Yield Stable Macroevolutionary

- Inferences? An Example from Squamate Reptiles. Systematic Biology, syw102.

 https://doi.org/10.1093/sysbio/syw102
- Treplin, S., Siegert, R., Bleidorn, C., Thompson, H. S., Fotso, R., & Tiedemann, R. (2008).

 Molecular phylogeny of songbirds (aves: Passeriformes) and the relative utility of

 common nuclear marker loci. *Cladistics*, 24(3), 328–349.
- Uyeda, J. C., Pennell, M. W., Miller, E. T., Maia, R., & McClain, C. R. (2017). The
 evolution of energetic scaling across the vertebrate tree of life. *The American*Naturalist, 190(2), 185–199.
- Vos, R. A., Balhoff, J. P., Caravas, J. A., Holder, M. T., Lapp, H., Maddison, W. P., ...
 others. (2012). NeXML: Rich, extensible, and verifiable representation of
 comparative data and metadata. *Systematic Biology*, 61(4), 675–689.
 https://doi.org/10.1093/sysbio/sys025
- Vos, R. A., & Mooers, A. Ø. (2004). Reconstructing divergence times for supertrees: A

 molecular approach. Phylogenetic Supertrees: Combining Information to Reveal the

 Tree of Life, 281–299.
- Webb, C. (2000). Exploring the Phylogenetic Structure of Ecological Communities: An

 Example for Rain Forest Trees. *The American Naturalist*, 156(2), 145–155.
- Webb, C., Ackerly, D., & Kembel, S. (2008). Phylocom: Software for the analysis of
 phylogenetic community structure and trait evolution. *Bioinformatics*, 24 (18),
 2098–2100. https://doi.org/10.1093/bioinformatics/btn358
- Webb, C., & Donoghue, M. (2005). Phylomatic: Tree assembly for applied phylogenetics.
 Molecular Ecology Notes, 5(1), 181–183.
- Weir, J., & Schluter, D. (2008). Calibrating the avian molecular clock. *Molecular Ecology*,

939 17(10), 2321-2328.

943

Zuccon, D., Prs-Jones, R., Rasmussen, P. C., & Ericson, P. G. (2012). The phylogenetic
 relationships and generic limits of finches (fringillidae). Molecular Phylogenetics and
 Evolution, 62(2), 581–596.

Figure Captions

Figure 1. DateLife's benchmarking results showing computation time used for taxon name processing and search across datelife's chronogram database, as a function of number of input taxon names (N). For each N = {10, 100, 200, ..., 1 000, ..., 9 000, 10 000}, we randomly sampled N species names from the class Aves, a hundred times, and then performed a datelife search processing the input names using the Taxon Names Resolution Service (TNRS; light gray), and without processing input names (dark gray). For comparison, we performed a chronogram search using names that have been pre-processed with TNRS.

Figure 2. DateLife results of an analysis of a small sample of 6 bird species within the
Passeriformes (a, b). Processed species names were found across 9 chronograms within 6
independent studies (c; Barker et al. (2012), Barker et al. (2015), Burns et al. (2014),
Hedges et al. (2015), Hooper and Price (2017), Jetz et al. (2012). This revealed 28 source
age data points for the queried species names (e; Table 1). Summarized age data (f; Table 2)
was used as secondary calibrations to date a tree topology obtained from OpenTree's
synthetic tree v13.4 (d), resulting in the chronogram of summary source ages shown in (g).

Figure 3. Median summary chronogram resulting from a Datelife analysis of bird species within the family Fringillidae. For visualization purposes, we are showing a portion of the final median summary chronogram encompassing 57 species out of the 289 total included in the analysis. The complete final chronogram is available as Supplementary Figure S3. The starting tree topology (Supplementary Fig. S2) has 289 tip species and 253 nodes; DateLife

revealed age data for 194 of these nodes from at least one published chronogram. In total, 19
different chronograms from 13 different studies contributed 818 age data points, which were
summarized to obtain a single value for each one of the 194 nodes with age data. From the
194 summary ages available, 21 were discarded and not used as calibrations (asterisk, *),
because they were older than a parent node or younger than a descendant node; the
remaining 173 summary ages were used as secondary calibrations (forward slash, /) with the
Branch Length Adjuster (BLADJ) software from Webb et al., (2008).

Figure 4. Cross validation of results from a DateLife analysis of the family Fringillidae,
shown in Figure 3 and Supplementary Figure S3. Each plot compares the original node age
estimates from an input source study chronogram (x axis) with the corresponding node age
resulting from a dating analysis using the DateLife workflow, excluding data from that study
(y axis).

- DateLife: leveraging databases and analytical tools to reveal the dated Tree of Life
- Luna L. Sánchez Reyes^{1,2}, Emily Jane McTavish¹, & Brian O'Meara²
 - ¹ University of California, Merced, USA
- ² University of Tennessee, Knoxville, USA

Author Note

- Department of Life and Environmental Sciences, University of California, Merced, CA 95343, USA.
- Department of Ecology and Evolutionary Biology, University of Tennessee, Knoxville, 8
- 446 Hesler Biology Building, Knoxville, TN 37996, USA.
- The authors made the following contributions. Luna L. Sánchez Reyes: Data curation, 10
- Investigation, Software, Visualization, Validation, Writing Original Draft Preparation, 11
- Writing Review & Editing; Emily Jane McTavish: Resources, Software, Writing Review & 12
- Editing; Brian O'Meara: Conceptualization, Funding acquisition, Methodology, Resources,
- Software, Supervision, Writing Review & Editing.
- Correspondence concerning this article should be addressed to Luna L. Sánchez Reyes, . 15 iC,
- E-mail: sanchez.reyes.luna@gmail.com

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

Abstract

17

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
- on of evolutiona.

 words: Tree; Phylogeny; S.

 affication; Supertree; Calibrations; .

 Word count: 6707-7484 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
- 52

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

76

77

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. 117

Searching a Chronogram Database

118

At the time of writing of this manuscript (Jun 11 Nov 21, 2023), DateLife's chronogram 119 database latest version consist of 253 chronograms published in 187 different studies, 120 encompassing close to 100 k (99 474) different species. It is curated from OpenTree's 121 phylogenetic database, the Phylogenetic database database, the Phylogenetic database 122 of expert and peer-reviewed phylogenetic knowledge with rich metadata and a wide 123 taxonomic scope (McTavish et al., 2015), which allows automatic and reproducible assembly of our chronogram database. Datelife's chronogram database is navigable as an R 125 data object within the datelife R package. We expect DateLife's database to largely 126 overlap with OpenTree's phylogenetic database taxonomic coverage, where Chordata and 127 Embryophyta are nearly fully sampled. In contrast, Bacteria, Fungi, Nematoda, and 128 Insecta, currently present a large gap between the number of named species and what has

phylogenetic information in OpenTree's synthetic tree. It is likely that users working with
the former groups will get results from a DateLife analysis. If none of the user's species are
found, the software will indicate the lack of age data for the queried taxa in the database.

A unique feature of the Phylesystem is that any user can add new published, 133 state-of-the-art chronograms any time, through OpenTree's curator application (https://tree.opentreeoflife.org/curator). Relying on an open source database permits an 135 automatic and reproducible assembly of DateLife's chronogram database, which is stored 136 and navigable as an R data object within the datelife R package. As chronograms are added to Phylesystem, they can be incorporated into the chronogram database of within the datelife R package, which is currently manually updated as new chronogram data is 139 added to Phylesystem by manually triggering an update. The updated datelife database 140 is assigned a new version number, followed by a package release on CRAN. We encourage 141 users to submit published chronograms to OpenTree's phylogenetic database, so that their 142 taxon of interest can be included in future DateLife searches. Users can directly run 143 datelife functions to trigger an update of their local chronogram database, to incorporate 144 any new chronograms to their DateLife analysis before an official a datelife database 145 update is released on CRAN. 146

A DateLife search is implemented by matching processed taxon names provided by the 147 user to tip labels in the chronogram database. Chronograms with at least two matching 148 taxon names on their tip labels are identified and pruned down to preserve only the matched 149 taxa. These matching pruned chronograms are referred to as source chronograms. Total distance in units of million years (Myr) between taxon pairs within each source chronogram 151 are stored as a patristic distance matrix (Fig. 1).... The matrix format speeds up extraction 152 of pairwise taxon ages of any queried taxa, as opposed to searching the ancestor node of a 153 pair of taxa in a "phylo" object or newick string. Finally, the patristic matrices are 154 associated to the study citation where the original chronogram was published, and stored as 155

an R object of the newly defined class datelifeResult.

157

161

162

163

164

165

166

167

171

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.
 - 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.
 - 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 172 users provide as input a tree topology from the literature, or one of their own making. If no 173 topology is provided, DateLife automatically extracts one from the OpenTree synthetic tree, 174 a phylogeny currently encompassing 2.3 million taxa across all life, assembled from 1,239 published phylogenetic trees and OpenTree's unified Taxonomy, OTT (Open Tree Of Life et 176 al., 2019). Alternatively, DateLife can combine topologies from source chronograms using a 177 supertree approach (Criscuolo, Berry, Douzery, & Gascuel, 2006). To do this, DateLife first 178 identifies the source chronograms that form a grove, roughly, a sufficiently overlapping set of 179 taxa between trees, by implementing definition 2.8 for n-overlap from Ané et al. (2009). If 180

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
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
combines topologies from source chronograms in a grove.

Applying Secondary Calibrations

Once a topology is chosen, DateLife applies the congruification method (Eastman, 187 Harmon, & Tank, 2013) that find nodes belonging to the same clade across source 188 chronograms, and then extracts the corresponding node ages from patristic distance matrices 189 stored as a datelifeResult object. Note that by definition, these matrices store total 190 distance (time from tip to tip), assuming that the terminal taxa are coeval and occur at the 191 present. Hence, node ages correspond to half the values stored in the datelifeResult 192 matrices. A table of congruified node ages that can be used as calibrations for a dating 193 analysis is stored as a congruifiedCalibrations object. 194

For each congruent node, the pairwise distances that traverse that node are summarized 195 into a single summary matrix using classic summary statistics (i.e., mean, median, minimum 196 and maximum ages), and the Supermatrix Distance Method (SDM; Criscuolo et al., 2006), 197 which deforms patristic distance matrices by minimizing variance and then averaging them. These single summary taxon pair age matrices are stored as summarized calibrations that can be used as secondary calibrations to date a tree topology - with or without initial branch lengths, using phylogenetic dating methods currently supported within DateLife: BLADJ 201 (Webb, Ackerly, & Kembel, 2008; Webb & Donoghue, 2005), MrBayes (Huelsenbeck & 202 Ronquist, 2001; Ronquist & Huelsenbeck, 2003), PATHd8 (Britton, Anderson, Jacquet, 203 Lundqvist, & Bremer, 2007), and treePL (Smith & O'Meara, 2012). 204

Dating a tree without branch lengths.—When producing or obtaining a tree 206 with branch lengths for a group of interest is not possible. DateLife can date a topology 207 without branch lengths, obtained from OpenTree or by implementing the supertree 208 approach described above, by applying the To date a tree topology when initial branch 209 lengths are unavailable, DateLife implements the Branch Length Adjuster (BLADJ-) 210 algorithm (Webb et al., 2008; Webb & Donoghue, 2005) algorithm, which requires no initial 211 branch lengths. The algorithm starts by fixing, which only requires a tree topology with 212 no branch lengths and at least two node ages to use as calibrations, one for the tree root 213 and one for any internal node of the topology. The BLADJ algorithm fixes ages for nodes 214 with calibration data upon the given tree topology. Then, it distributes time for assigns ages 215 to nodes with no data available age information by distributing time evenly between 216 calibrated nodes, minimizing age variance in the resulting chronogram. This approach has 217 proven useful for ecological analyses that require a phylogenetic time context (Webb et al., 218 2008). When there is conflict between ages of calibrated nodes, BLADJ ignores node ages 219 that are older than the age of a parent node. The BLADJ algorithm requires a root age 220 estimate to run. If-Users can provide an appropriate root age estimate of their own or one 221 obtained from the 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 provided, DateLife will choose an DateLife chooses a random age for 224 the rootso that it can return a dated topology. It will also, so that a dated tree topology 225 can be generated with BLADJ. In this case, DateLife will provide a conspicuous warning 226 message, so that users are aware that the root of the chronogram does not have information 227 available was chosen at random because there was no information available for it in the 228 chronogram database, along with suggestions on how the user can find and provide an 229 appropriate age for the root of the initial topology. 230

In the absence of genetic data, BLADJ is a very agnostic way to assign ages to nodes with no available data, as it does not require any assumptions on the underlying model of

branch length distribution. It is however An alternative to BLADJ to date tree topologies in the absence of initial branch lengths that is common practice in the literature is to use a 234 birth-death model to assign ages to nodes with no genetic data draw branch lengths (Jetz, 235 Thomas, Joy, Hartmann, & Mooers, 2012; Rabosky et al., 2018; Smith & Brown, 2018). To 236 do so, DateLife implements In addition to the initial tree topology and nodes with age 237 data, these methods require initial values of speciation and extinction rate parameters 238 provided by the user. DateLife implements this approach with MrBayes (Huelsenbeck & 230 Ronquist, 2001; Ronquist & Huelsenbeck, 2003), using nodes with published age data as 240 calibration priors on a fixed topology nodes of a tree topology with no branch lengths, a 241 simple birth-death model with parameters that can be determined speciation and extinction 242 rate parameters that are provided by the user, and no genetic data. However, BLADJ is the 243 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. 245

Dating a tree with branch lengths.—Relative branch lengths can provide key 246 information for phylogenetic dating, specifically for nodes without any calibration data 247 available. While using initial branch length data is the golden standard for phylogenetic 248 dating analyses, estimating trees with branch lengths proportional to substitution rates per 249 site obtaining such information from scratch is not an easy task; it requires obtaining 250 primary data, assembling and curating a homology (orthology) hypothesis, and choosing and 251 implementing a method for phylogenetic inference. DateLife implements a workflow to 252 streamline this process by applying open data from the Barcode of Life Data System, BOLD 253 (Ratnasingham & Hebert, 2007) to obtain genetic markers for input taxa. By default, BOLD 254 genetic sequences are aligned with MUSCLE (Edgar, 2004) using functions from the msa R 255 package (Bodenhofer et al., 2015). Alternatively, sequences can be aligned with MAFFT (Katoh, Asimenos, & Toh, 2009), using functions from the ape R package (Paradis et al., 257 2004). The BOLD sequence alignment is then used to obtain initial branch lengths with the 258 accelerated transformation (ACCTRAN) parsimony algorithm, which resolves ambiguous 250

character optimization by assigning changes along branches of the tree as close to the root as 260 possible (Agnarsson & Miller, 2008), resulting in older internal nodes as compared to other 261 parsimony algorithms (Forest et al., 2005). The parsimony branch lengths are then 262 optimized using Maximum Likelihood, given the alignment, the topology and a simple 263 Jukes-Cantor model, producing a BOLD tree with branch lengths proportional to expected 264 number of substitutions per site. Both parsimony and ML optimizations are done with 265 functions from the phangorn package (Schliep, 2011). Due to the computing load it requires, 266 the BOLD workflow is currently only supported through DateLife's R package. It is not yet 267 available through the web application. 268

Phylogenetic dating methods supported in DateLife that incorporate branch length information from the input topology in combination with the secondary calibrations include:
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 method (Sanderson, 2002); and MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), a Bayesian inference program implementing Markov chain Monte Carlo (MCMC) methods to estimate a posterior distribution of model parameters.

Visualizing Results

Finally, users can save all source and summary chronograms in formats 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).

Benchmark

276

282

R package datelife code speed was tested on an Apple iMac with one 3.4 GHz Intel
Core i5 processor. We registered variation in computing time of query processing and search

308

309

through the database relative to number of queried taxon names. 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 ten thousand names can be processed and searched in less than 30 minutes with the most time consuming settings. Once names have been processed as described in methods, a name search through the chronogram database can be performed in less than a minute, even with a very large number of taxon names (Fig. 21).

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

We illustrate the DateLife workflow using a family within the passeriform birds 300 encompassing the true finches, Fringillidae, as case study. On a small example, we analysed 6 301 bird species, and results from each step of the workflow are shown in Figure 3.—2. As a 302 second example, we analysed 289 bird species in the family Fringillidae that are included in 303 the NCBI taxonomy. The summary chronogram resulting. One clade from the full summary 304 chronogram result from the DateLife analysis is shown in Figure 5, and Figure 3. The full 305 chronogram for all 289 species and the results from previous steps of the workflow are 306 available as Supplementary Figures. 307

A Small Example

Creating a search query.— We chose 6 bird species within the Passeriformes. The

sample includes two species of cardinals: the black-thighed grosbeak – Pheucticus tibialis 310 and the crimson-collared grosbeak – Rhodothraupis celaeno; three species of buntings: the 311 yellowhammer – Emberiza citrinella, the pine bunting – Emberiza leucocephalos and the 312 yellow-throated bunting – Emberiza elegans; and one species of tanager, the vegetarian finch – 313 Platyspiza crassirostris. Processing of input names found that Emberiza elegans is synonym 314 for Schoeniclus elegans in the default reference taxonomy (OTT v3.3, June 1, 2021). For a 315 detailed discussion on the state of the synonym, refer to Avibase (Avibase, 2022; Lepage, 316 2004; Lepage, Vaidya, & Guralnick, 2014). Discovering this synonym allowed assigning five 317 age data points for the parent node of Emberiza elegans, shown as Schoeniclus elegans in 318 Figure 3a2, which would not have had any data otherwise. 319

Searching the database. - DateLife used the processed input names to search the 320 local chronogram database and found 9 matching chronograms from 6 different studies (Fig. 321 3b2c). Three studies matched five input names (Barker, Burns, Klicka, Lanyon, & Lovette, 322 2015; Hedges, Marin, Suleski, Paymer, & Kumar, 2015; Jetz et al., 2012), one study matched 323 four input names (Hooper & Price, 2017) and two studies matched two input names (Barker, 324 Burns, Klicka, Lanyon, & Lovette, 2013; Burns et al., 2014). No studies matched all input 325 names. Together, source chronograms provide 28 unique age data points, covering all nodes 326 on our chosen tree topology to date (Table 1). 327

Summarizing search results.— DateLife obtained OpenTree's synthetic tree
topology for these taxa (Fig. 3e2d), and congruified and mapped age data to nodes in this
chosen topology, shown in Table 1. The name processing step allowed including five data
points for node "n4" (parent of Schoeniclus elegans; Fig. 3A) that would not have had any
data otherwise due to name mismatch. Age summary statistics per node were calculated
(Table 2) and used as calibrations to date the tree topology using the BLADJ algorithm. As
expected, more inclusive nodes (e.g., node "n1") have more variance in age data than less
inclusive nodes (e.g., node "n5"). Median summary age data for node "n2" was excluded as

final calibration because it is older than the median age of a more inclusive node, "n1" (Fig.

337 3c42g).



An Example with the Family of True Finches

Creating a query. To obtain ages for all species within the family of true finches, 339 Fringillidae, we ran a DateLife query using the "get species from taxon" flag, which gets all 340 recognized species names within a named group from a taxonomy of choice. Following the 341 NCBI taxonomy, our DateLife query has 289 Fringillidae species names. This 342 taxon-constrained approach implies that the full DateLife analysis will be performed using a 343 tree topology and ages available for species names from a given taxonomic group, which do 344 not necessarily correspond to a monophyletic group. Users can change this behavior by 345 providing all species names corresponding to a monophyletic group as input for a DateLife 346 search, or a monophyletic tree to construct a DateLife summary. 347

Searching the database. Next, we used the processed species names in our 348 DateLife query to identify chronograms with at least two Fringillidae species as tip taxa. 340 The DateLife search identified 19 chronograms matching this criteria, published in 13 350 different studies (Barker et al., 2013, 2015; Burns et al., 2014; Claramunt & Cracraft, 2015; 351 Gibb et al., 2015; Hedges et al., 2015; Hooper & Price, 2017; Jetz et al., 2012; Kimball et al., 352 2019; Oliveros et al., 2019; Price et al., 2014; Roquet, Lavergne, & Thuiller, 2014; Uyeda, 353 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 368 congruified to OpenTree's topology shown in Figure 4bSupplementary Figure S2, reducing 369 the original 1,206 node age data set to 818 different data points (Supplementary Table S2) 370 that can could be used as calibrations for the chosen topology (Fig. 4b). that chosen 371 topology. The congruent node age data points were summarized for each node, resulting in 372 194 summary node ages. From these 21 were excluded as secondary calibrations because they 373 were older than the ancestral node. The remaining 173 summary node ages were used as 374 secondary calibrations to obtain a fully dated (and resolved) phylogeny with the program 375 BLADJ (Fig. 5). Supplementary Figure S3). Results for a subgroup are shown in Figure 3. 376

CROSS-VALIDATION TEST

We performed a cross validation test of a DateLife analysis using the Fringillidae 378 source chronograms obtained above (Supplementary Figs. S2-S20). We used as S5-S23). As 379 inputs for a DateLife analysis, we used all individual tree topologies from each of the 19 380 source chronograms from 13 studies, treating their node ages as unknown. We congruified 381 node ages extracted from chronograms from all other studies upon the individual topologies, 382 effectively excluding original ages from each topology. Finally, average node ages per node 383 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 385 available are largely generally correlated (Fig. 64). For five studies, DateLife tended to 386 underestimate ages for topologically deeper nodes (those with many descendant taxa, aka 387 "closer to the root") relative to the original estimate, and overestimate ages for nodes closer 388 to the tips. Accordingly, root ages are generally older in the original study than estimated 389

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.

393 DISCUSSION

DateLife's goal is to improve availability and accessibility, accessibility, and reusability
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, use 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
that users want to work with, with—to use any of its functionalities.

A total of 99,474 unique terminal taxa are represented in DateLife's database. 400 Incorporation of more chronograms into the database will continue to improve DateLife's 401 services. One option to increase the number of chronograms in the DateLife database is the 402 Dryad data repository. Methods to automatically mine chronograms from Dryad could be 403 designed and implemented. However, Dryad's metadata system has no information to 404 automatically detect branch length units, and those would still need to be determined 405 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 407 OpenTree's Phylesystem (McTavish et al., 2015) for the benefit of research and the scientific 408 community as a whole. Another important source of expert data on time of lineage divergence is TimeTree's database (Hedges, Dudley, & Kumar, 2006), which holds chronograms from more than 4k published studies, and is fully browsable using its graphical user interface (timetree.org). TimeTree's chronogram database was not accessible in computer readable format until very recently (Kumar et al., 2022), when its terms of use 413 and website application were updated, now allowing some kinds of reuse, but not 414 redistribution. 415

417

418

419

420

438

The inaccessibility of TimeTree's database was an inspiration for the DateLife project, which was born as a prototype tool initially developed over a series of hackathons at the National Evolutionary Synthesis Center, NC, USA (Stoltzfus et al., 2013), as the need to make scientific information that is funded by the public practically available to the public was acknowledged and prioritized.

As we envision that DateLife will have many interesting applications in research and 421 beyond, we emphasize that DateLife's results –as well as any insights gleaned from them, 422 largely depend on the quality of the source chronograms: low quality chronograms will 423 produce low quality results. The "garbage in, garbage out" problem has long been recognised 424 in supertree methods for summarizing phylogenetic trees (Bininda-Emonds et al., 2004). We 425 note that this is a surfacing issue of any automated tool for biological data analysis. For 426 example, DNA riddled with sequencing errors will produce generally poor alignments that 427 will return biased evolutionary hypothesis, independently of the quality of the analysis 428 software used. Again, we urge readers and DateLife users to explore all input chronograms before using a summary chronogram resulting from a DateLife workflow. 430

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 the best 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, for example, the robust ongoing debate about crown group age of angiosperms

(Barba-Montoya, Reis, Schneider, Donoghue, & Yang, 2018; Magallón, Gómez-Acevedo, Sánchez-Reyes, & Hernández-Hernández, 2015; Ramshaw et al., 1972; Sanderson & Doyle, 442 2001; Sauquet, Ramírez-Barahona, & Magallón, 2021). Alternative source chronograms 443 available for the same taxa have potentially been estimated implementing different types of 444 calibrations, which affects the resulting node age estimates. For example, in the DateLife 445 analysis of the Fringillidae shown above, the chronograms from one study (Burns et al., 446 2014) were inferred using molecular substitution rate estimates across birds (Weir & 447 Schluter, 2008), and have much older age estimates for the same nodes than chronograms that were inferred using fossil calibrations (Figs. 5, 6; 3, 4c, Supplementary Figs. S1c, S4S4c, 440 S10). Another source of conflict in estimated node ages can arise from different placements 450 for the same calibration, which would imply fundamentally distinct evolutionary hypotheses 451 (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 placement. Yet, if 453 one researcher concludes that a fossil should calibrate the ingroup of a clade, while another researcher concludes that the same fossil should calibrate the outgroup of the clade, the 455 resulting age estimates will differ, as the placement of calibrations as stem or crown group is 456 known to significantly affect estimates of time of lineage divergence (Sauquet, 2013). Finally, 457 placement of calibrations also affects uncertainty of node age estimates. For example, nodes 458 that are sandwiched between a calibrated node and a calibrated root have less freedom of 459 movement and hence narrower confidence intervals (Vos & Mooers, 2004), which inflates 460 precision for nodes without calibrations but does not necessarily improve accuracy of the 461 estimated ages. 462

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

DateLife constructs summary chronograms using node ages extracted from existing 470 chronograms, i.e. secondary calibrations. In general, the scientific community has more 471 confidence in chronograms using primary calibrations, where the dated tree is generated from 472 a single analysis where carefully chosen fossil calibrations are the source of absolute time 473 information, than in analyses dated using secondary calibrations (Antonelli et al., 2017; 474 Garzón-Orduña, Silva-Brandão, Willmott, Freitas, & Brower, 2015; Graur & Martin, 2004; 475 Sauguet, 2013; Sauguet et al., 2012; Schenk, 2016; Shaul & Graur, 2002). However, 476 implementation of primary calibrations is difficult: it requires specialized expertise and 477 training to discover, place and apply calibrations appropriately (Hipsley & Müller, 2014; 478 Ksepka et al., 2011). One approach is to use fossils that have been widely discussed and 479 previously curated as calibrations to date other trees (Ksepka et al., 2011; Sauquet, 2013), 480 and making sure that all data reflect a coherent evolutionary history (Sauquet, 2013), as for 481 example done by Antonelli et al. (2017). The Fossil Calibration Database provides data for 482 220 primary calibration points encompassing flowering plants and metazoans, that have been 483 curated by experts and used for dating analysis in peer-reviewed publications (Ksepka et al., 2015). This database facilitates the use of expert primary fossil calibrations in new 485 phylogenetic dating analyses. Yet, users still require the expertise to locate and calibrate appropriate nodes in their phylogenies which correspond with fossils available in the database.

Recently, Powell, Waskin, and Battistuzzi (2020) showed in a simulation study that
secondary calibrations using node ages based on previous molecular clock analyses can be as
good as primary calibrations. Using several secondary calibrations (as opposed to just one)
can provide sufficient information to alleviate or even neutralize potential biases (Graur &
Martin, 2004; Sauquet, 2013; Shaul & Graur, 2002). Our cross validation analysis also

provides insight into the application of secondary calibrations. Node ages summarized with 494 DateLife and those from the original studies are well correlated (Supplementary Figs. 495 S2-S20S5-S23). We also note that DateLife estimates for nodes closer to the root tend to be 496 slightly younger than ages from the original studies. In contrast, nodes closer to the tips 497 tend to be slightly older when estimated using our secondary calibrations than ages from the 498 original studies. The only exception to this trend was observed in Burns et al. (2014) 490 chronogram, which generally displays much younger node ages when estimated using 500 secondary calibrations than the original study (Supplementary Figs. S1, S5S4c, S10), 501 supporting previous observations (Sauquet et al., 2012; Schenk, 2016). However, these 502 younger dates are more likely an example of how multiple secondary calibrations can correct 503 erroneous estimates, as dates on the Burns et al. (2014) tree were obtained using a single 504 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. 5).

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, 20023, Supplementary Fig.
S3).

Sumarizing Chronograms

512

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.

528

529

530

531

543

544

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 is a common 532 resource in research, as it provides the opportunity to work with a chronogram that reflects 533 a unified evolutionary history for a lineage, by putting together evidence from different 534 hypotheses. The For example, the largest, and taxonomically broadest chronogram currently available from OpenTree was constructed summarizing age data from 2,274 published chronograms using NCBI's taxonomic tree as backbone (Hedges et al., 2015). A 537 summarizing exercise may also, which has been widely reused for research. Finally, we 538 note that summarizing chronograms should be done with caution, as it may amplify the 530 effect of uncertainty and errors in source data, and blur parts of the evolutionary history of a lineage that might only be reflected in source chronograms and lost on the summary 541 chronogram (Sauguet et al., 2021). 542

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 552 backbone taxonomy is a common practice in scientific publications: Jetz et al. (2012), 553 created a chronogram of all 9,993 bird species, where 67% had molecular data and the rest 554 was simulated; Rabosky et al. (2018) created a chronogram of 31,536 ray-finned fishes, of 555 which only 37% had molecular data; Smith and Brown (2018) constructed a chronogram of 556 353,185 seed plants where only 23% had molecular data. These stochastically resolved 557 chronograms Stochastically resolved chronograms can return diversification rates estimates 558 that appear less biased than those estimated from their incompletely sampled counterparts, 559 even with methods that account for missing lineages by using sampling fractions (Chang et 560 al., 2020; Cusimano et al., 2012), but can also introduce spurious patterns of early bursts of 561 diversification (Cusimano & Renner, 2010; Sun et al., 2020). 562

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 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 cannot be accurately inferred with phylogenies that have been resolved stochastically over a

592

594

595

taxonomic backbone, as any patterns would be erased by randomization (Rabosky, 2015). 571 We note that the same applies for geography- and morphology-dependent diversification 572 analysis. Hence, we suggest that phylogenies that have been processed with taxonomy-based 573 stochastic polytomy resolvers, including certain summary chronograms from a DateLife 574 analysis, can be useful as null or neutral models, representing the case of a diversification 575 process that is independent of traits and geographical scenario. 576

Taxonomy-based stochastic polytomy resolvers have been used to advance research in 577 evolution, still, risks come with this practice. Taken to the extreme, one could generate a 578 fully resolved, calibrated tree of all modern and extinct taxa using a single taxonomy, a 579 single calibration, and assigning branch lengths following a birth-death diversification model. 580 Clearly, this can lead to a misrepresentation of the true evolutionary history. We urge 581 DateLife users to follow the example of the large tree papers cited above, by carefully 582 considering the statistical assumptions being made, potential biases, and assessing the 583 consistency of DateLife's results with prior work.

Conclusions

Knowledge of the evolutionary time frame of organisms is key to many research areas: 586 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.

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

616

web-based R shiny application at www.datelife.org. DateLife provides a straightforward way 596 to get an informed picture of the state of knowledge of for the time frame of evolution of 597 different regions of the tree of life, and allows identifying regions that require more research, 598 or that have conflicting information. Additionally, both summary and newly generated trees 599 using the DateLife workflow are useful to evaluate evolutionary hypotheses in different areas 600 of research. We hope that the DateLife project will increase awareness of the existing 601 variation in expert estimations of time of divergence, and foster exploration of the effect of 602 alternative divergence time hypotheses on the results of analyses, nurturing a culture of more 603 cautious interpretation of evolutionary results. 604

AVAILABILITY

The DateLife software is free and open source. It can be used online through its R 606 shiny web application hosted at http://www.datelife.org, and locally through the datelife 607 R package, available from Zenodo (https://doi.org/10.5281/zenodo.593938 and the CRAN 608 repository (Sanchez-Reyes et al., 2022). DateLife's web application is maintained using 609 RStudio's shiny server and the shiny package open infrastructure, as well as Docker and 610 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 613 repository (https://github.com/phylotastic/datelife) and can be installed using the 614 command devtools::install github("phylotastic/datelife"). 615

Supplementary Material

Supplementary Figures can be viewed and downloaded from their Zenodo repository

(). Supplementary material, including figures, tables, code, biological examples, benchmark

results, data files and online-only appendices, can be viewed and downloaded from the

Dryad data repository (https://doi.org/10.5061/dryad.cnp5hqc6w), as well as in-from the

- Zenodo stable repositories that host the reproducible manuscript
- 622 (https://doi.org/10.5281/zenodo.7435094), the biological examples
- 623 (https://doi.org/10.5281/zenodo.7435101), and the software benchmark
- 624 (https://doi.org/10.5281/zenodo.7435106)—, and the figures
- 625 (https://doi.org/10.5281/zenodo.6683667). Dryad's data publication fee is covered by the
- Society of Systematic Biologists. Development versions corresponding to all of the above are
- hosted on GitHub, accessible accessible at https://github.com/LunaSare/datelifeMS1,
- 628 https://github.com/LunaSare/datelife examples, and
- 629 https://github.com/LunaSare/datelife_benchmark.

630 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 reviewer,
for comments that greatly improved this manuscript. The DateLife project was born as a
prototype tool aiming to provide 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 Lab at the University
of Tennessee Knoxville for suggestions, discussions and software testing. The late National
Evolutionary Synthesis Center (NESCent), which sponsored hackathons that led to initial
work on this project. The team that assembled DateLife's first proof of concept: Tracy
Heath, Jonathan Eastman, Peter Midford, Joseph Brown, Matt Pennell, Mike Alfaro, and
Luke Harmon. The Open Tree of Life project that provides the open, metadata rich

repository of trees used to construct DateLife's chronogram database. The many scientists
who publish their chronograms in an open, reusable form, and the scientists who curate them
for deposition in the Open Tree of Life repository. The NSF for funding nearly all the above,
in addition to the ABI grant that funded this project itself.



References

- Agnarsson, I., & Miller, J. A. (2008). Is ACCTRAN better than DELTRAN? Cladistics,
 24 (6), 1032–1038.
- Alström, P., Hooper, D. M., Liu, Y., Olsson, U., Mohan, D., Gelang, M., ... Price, T. D.

 (2014). Discovery of a relict lineage and monotypic family of passerine birds. *Biology Letters*, 10(3), 20131067.
- Ané, C., Eulenstein, O., Piaggio-Talice, R., & Sanderson, M. J. (2009). Groves of phylogenetic trees. *Annals of Combinatorics*, 13(2), 139–167.
- Antonelli, A., Hettling, H., Condamine, F. L., Vos, K., Nilsson, R. H., Sanderson, M. J., ...

 Vos, R. A. (2017). Toward a self-updating platform for estimating rates of speciation
 and migration, ages, and relationships of Taxa. Systematic Biology, 66(2), 153–166.

 https://doi.org/10.1093/sysbio/syw066
- Archie, J., Day, W. H., Felsenstein, J., Maddison, W., Meacham, C., Rohlf, F. J., & Swofford, D. (1986). The Newick tree format. Retrieved from {https://evolution.genetics.washington.edu/phylip/newicktree.html}
- Avibase. (2022). Yellow-throated Bunting. Avibase the World Bird Database, (Online Resource). Retrieved from {https://avibase.bsc-eoc.org/species.jsp?lang=EN&avibaseid=82D1EE0049D8D927}
- Bapst, D. W. (2012). Paleotree: An R package for paleontological and phylogenetic analyses of evolution. *Methods in Ecology and Evolution*, 3(5), 803–807.
- https://doi.org/10.1111/j.2041-210X.2012.00223.x
- Barba-Montoya, J., Reis, M. dos, Schneider, H., Donoghue, P. C., & Yang, Z. (2018).
- Constraining uncertainty in the timescale of angiosperm evolution and the veracity of

- a cretaceous terrestrial revolution. New Phytologist, 218(2), 819–834.
- Barker, F. K. (2014). Mitogenomic data resolve basal relationships among passeriform and passeridan birds. *Molecular Phylogenetics and Evolution*, 79, 313–324.
- Barker, F. K., Burns, K. J., Klicka, J., Lanyon, S. M., & Lovette, I. J. (2013). Going to
 extremes: Contrasting rates of diversification in a recent radiation of new world
 passerine birds. Systematic Biology, 62(2), 298–320.
- Barker, F. K., Burns, K. J., Klicka, J., Lanyon, S. M., & Lovette, I. J. (2015). New insights into new world biogeography: An integrated view from the phylogeny of blackbirds, cardinals, sparrows, tanagers, warblers, and allies. *The Auk: Ornithological Advances*, 132(2), 333–348.
- Barker, F. K., Cibois, A., Schikler, P., Feinstein, J., & Cracraft, J. (2004). Phylogeny and diversification of the largest avian radiation. *Proceedings of the National Academy of Sciences*, 101(30), 11040–11045.
- Beresford, P., Barker, F., Ryan, P., & Crowe, T. (2005). African endemics span the tree of songbirds (passeri): Molecular systematics of several evolutionary "enigmas".

 Proceedings of the Royal Society B: Biological Sciences, 272(1565), 849–858.
- Bininda-Emonds, O. R., Jones, K. E., Price, S. A., Cardillo, M., Grenyer, R., & Purvis, A.
 (2004). Garbage in, garbage out: Data issues in supertree construction. *Phylogenetic Supertrees: Combining Information to Reveal the Tree of Life*, 267–280.
- Bodenhofer, U., Bonatesta, E., Horejš-Kainrath, C., & Hochreiter, S. (2015). Msa: An r
 package for multiple sequence alignment. *Bioinformatics*, 31(24), 3997–3999.
- Boyle, B., Hopkins, N., Lu, Z., Raygoza Garay, J. A., Mozzherin, D., Rees, T., ... Enquist,
 B. J. (2013). The taxonomic name resolution service: An online tool for automated

- standardization of plant names. BMC Bioinformatics, 14(1).
- 697 https://doi.org/10.1186/1471-2105-14-16
- ⁶⁹⁸ Britton, T., Anderson, C. L., Jacquet, D., Lundqvist, S., & Bremer, K. (2007). Estimating
- Divergence Times in Large Phylogenetic Trees. Systematic Biology, 56 (788777878),
- 700 741–752. https://doi.org/10.1080/10635150701613783
- Bryson Jr, R. W., Chaves, J., Smith, B. T., Miller, M. J., Winker, K., Pérez-Emán, J. L., &
- Klicka, J. (2014). Diversification across the new world within the 'blue' cardinalids
- (aves: Cardinalidae). Journal of Biogeography, 41(3), 587–599.
- Burleigh, J. G., Kimball, R. T., & Braun, E. L. (2015). Building the avian tree of life using a
- large-scale, sparse supermatrix. Molecular Phylogenetics and Evolution, 84, 53–63.
- Burns, K. J., Shultz, A. J., Title, P. O., Mason, N. A., Barker, F. K., Klicka, J., ... Lovette,
- I. J. (2014). Phylogenetics and diversification of tanagers (passeriformes:
- Thraupidae), the largest radiation of neotropical songbirds. *Molecular Phylogenetics*
- and Evolution, 75, 41–77.
- Chamberlain, S. (2018). bold: Interface to Bold Systems API. Retrieved from
- https://CRAN.R-project.org/package=bold
- ⁷¹² Chamberlain, S. A., & Szöcs, E. (2013). taxize: taxonomic search and retrieval in R [version
- 2; referees: 3 approved]. F1000Research, 2(191), 1–29.
- https://doi.org/10.12688/f1000research.2-191.v2
- ⁷¹⁵ Chang, J., Rabosky, D. L., & Alfaro, M. E. (2020). Estimating diversification rates on
- incompletely sampled phylogenies: Theoretical concerns and practical solutions.
- Systematic Biology, 69(3), 602–611.
- Chaves, J. A., Hidalgo, J. R., & Klicka, J. (2013). Biogeography and evolutionary history of

- the n eotropical genus s altator (a ves: T hraupini). Journal of Biogeography, 40(11), 2180–2190.
- Claramunt, S., & Cracraft, J. (2015). A new time tree reveals earth history's imprint on the evolution of modern birds. *Science Advances*, 1(11), e1501005.
- Criscuolo, A., Berry, V., Douzery, E. J., & Gascuel, O. (2006). SDM: A fast distance-based approach for (super)tree building in phylogenomics. Systematic Biology, 55(5), 740–755. https://doi.org/10.1080/10635150600969872
- Cusimano, N., & Renner, S. S. (2010). Slowdowns in diversification rates from real phylogenies may not be real. *Systematic Biology*, 59(4), 458–464.
- Cusimano, N., Stadler, T., & Renner, S. S. (2012). A new method for handling missing
 species in diversification analysis applicable to randomly or nonrandomly sampled
 phylogenies. Systematic Biology, 61(5), 785–792.
- Delsuc, F., Philippe, H., Tsagkogeorga, G., Simion, P., Tilak, M.-K., Turon, X., . . . Douzery,
 E. J. (2018). A phylogenomic framework and timescale for comparative studies of
 tunicates. *BMC Biology*, 16(1), 1–14.
- Eastman, J. M., Harmon, L. J., & Tank, D. C. (2013). Congruification: Support for time scaling large phylogenetic trees. *Methods in Ecology and Evolution*, 4(7), 688–691. https://doi.org/10.1111/2041-210X.12051
- Edgar, R. C. (2004). MUSCLE: Multiple sequence alignment with high accuracy and high throughput. *Nucleic Acids Research*, 32(5), 1792–1797.
- Felsenstein, J. (1985). Phylogenies and the Comparative Method. *The American Naturalist*, 125(1), 1–15. Retrieved from http://www.jstor.org/stable/2461605
- Forest, F., Savolainen, V., Chase, M. W., Lupia, R., Bruneau, A., & Crane, P. R. (2005).

- Teasing apart molecular-versus fossil-based error estimates when dating phylogenetic trees: A case study in the birch family (betulaceae). Systematic Botany, 30(1), 118–133.
- Freckleton, R. P., Harvey, P. H., & Pagel, M. (2002). Phylogenetic analysis and comparative data: A test and review of evidence. *The American Naturalist*.
- Garzón-Orduña, I. J., Silva-Brandão, K. L., Willmott, K. R., Freitas, A. V., & Brower, A. V.

 (2015). Incompatible ages for clearwing butterflies based on alternative secondary

 calibrations. Systematic Biology, 64(5), 752–767.
- GBIF Secretariat. (2022). GBIF Backbone Taxonomy. *Checklist dataset*, (Online Resource accessed via GBIF.org). Retrieved from {https://doi.org/10.15468/39omei }
- Gibb, G. C., England, R., Hartig, G., McLenachan, P. A., Taylor Smith, B. L., McComish,
 B. J., ... Penny, D. (2015). New zealand passerines help clarify the diversification of
 major songbird lineages during the oligocene. *Genome Biology and Evolution*, 7(11),
 2983–2995.
- Graur, D., & Martin, W. (2004). Reading the entrails of chickens: Molecular timescales of evolution and the illusion of precision. *TRENDS in Genetics*, 20(2), 80–86.
- Hackett, S. J., Kimball, R. T., Reddy, S., Bowie, R. C., Braun, E. L., Braun, M. J., ...
 others. (2008). A phylogenomic study of birds reveals their evolutionary history.

 Science, 320(5884), 1763–1768.
- Harvey, P. H., Pagel, M. D., & others. (1991). The comparative method in evolutionary
 biology (Vol. 239). Oxford university press Oxford.
- Hedges, S. B., Dudley, J., & Kumar, S. (2006). TimeTree: A public knowledge-base of divergence times among organisms. *Bioinformatics*, 22(23), 2971–2972.

- https://doi.org/10.1093/bioinformatics/btl505
- Hedges, S. B., Marin, J., Suleski, M., Paymer, M., & Kumar, S. (2015). Tree of life reveals
 clock-like speciation and diversification. *Molecular Biology and Evolution*, 32(4),
 835–845. https://doi.org/10.1093/molbev/msv037
- Heibl, C. (2008). PHYLOCH: R language tree plotting tools and interfaces to diverse

 phylogenetic software packages. Retrieved from

 http://www.christophheibl.de/Rpackages.html
- Hipsley, C. A., & Müller, J. (2014). Beyond fossil calibrations: Realities of molecular clock practices in evolutionary biology. *Frontiers in Genetics*, 5, 138.
- Hooper, D. M., & Price, T. D. (2017). Chromosomal inversion differences correlate with range overlap in passerine birds. *Nature Ecology & Evolution*, 1(10), 1526.
- Huelsenbeck, J. P., & Ronquist, F. (2001). MRBAYES: Bayesian inference of phylogenetic
 trees. Bioinformatics, 17(8), 754–755.
 https://doi.org/10.1093/bioinformatics/17.8.754
- Jetz, W., Thomas, G., Joy, J. J., Hartmann, K., & Mooers, A. (2012). The global diversity
 of birds in space and time. *Nature*, 491 (7424), 444–448.
- 781 https://doi.org/10.1038/nature11631
- Johansson, U. S., Fjeldså, J., & Bowie, R. C. (2008). Phylogenetic relationships within

 passerida (aves: Passeriformes): A review and a new molecular phylogeny based on

 three nuclear intron markers. *Molecular Phylogenetics and Evolution*, 48(3), 858–876.
- Katoh, K., Asimenos, G., & Toh, H. (2009). Multiple alignment of dna sequences with mafft.

 In *Bioinformatics for dna sequence analysis* (pp. 39–64). Springer.
- Kimball, R. T., Oliveros, C. H., Wang, N., White, N. D., Barker, F. K., Field, D. J., ...

- others. (2019). A phylogenomic supertree of birds. *Diversity*, 11(7), 109.
- Klicka, J., Barker, F. K., Burns, K. J., Lanyon, S. M., Lovette, I. J., Chaves, J. A., & Bryson
- Jr, R. W. (2014). A comprehensive multilocus assessment of sparrow (aves:
- Passerellidae) relationships. Molecular Phylogenetics and Evolution, 77, 177–182.
- Ksepka, D. T., Benton, M. J., Carrano, M. T., Gandolfo, M. A., Head, J. J., Hermsen, E. J.,
- ... others. (2011). Synthesizing and databasing fossil calibrations: Divergence dating
- and beyond. The Royal Society.
- Ksepka, D. T., Parham, J. F., Allman, J. F., Benton, M. J., Carrano, M. T., Cranston, K.
- A., ... others. (2015). The fossil calibration database—a new resource for divergence
- dating. Systematic Biology, 64(5), 853-859.
- Kumar, S., Suleski, M., Craig, J. M., Kasprowicz, A. E., Sanderford, M., Li, M., ...
- Hedges, S. B. (2022). TimeTree 5: An expanded resource for species divergence
- times. Molecular Biology and Evolution, 39(8), msac174.
- Lamichhaney, S., Berglund, J., Almén, M. S., Maqbool, K., Grabherr, M., Martinez-Barrio,
- A., ... others. (2015). Evolution of darwin's finches and their beaks revealed by
- genome sequencing. *Nature*, 518(7539), 371–375.
- Laubichler, M. D., & Maienschein, J. (2009). Form and function in developmental evolution.
- 805 Cambridge University Press.
- Lepage, D. (2004). Avibase: The world bird database. Bird Studies Canada.
- Lepage, D., Vaidya, G., & Guralnick, R. (2014). Avibase—a database system for managing
- and organizing taxonomic concepts. ZooKeys, (420), 117.
- Lerner, H. R., Meyer, M., James, H. F., Hofreiter, M., & Fleischer, R. C. (2011). Multilocus
- resolution of phylogeny and timescale in the extant adaptive radiation of hawaiian

- honeycreepers. Current Biology, 21 (21), 1838–1844.
- Lovette, I. J., Pérez-Emán, J. L., Sullivan, J. P., Banks, R. C., Fiorentino, I.,
- Córdoba-Córdoba, S., ... others. (2010). A comprehensive multilocus phylogeny for
- the wood-warblers and a revised classification of the parulidae (aves). *Molecular*
- Phylogenetics and Evolution, 57(2), 753-770.
- Magallon, S., & Sanderson, M. (2001). Absolute diversification rates in angiosperm clades.
- Evolution, 55(9), 1762-1780.
- Magallón, S. (2010). Using fossils to break long branches in molecular dating: A comparison
- of relaxed clocks applied to the origin of angiosperms. Systematic Biology, 59(4),
- 384–399
- Magallón, S., Gómez-Acevedo, S., Sánchez-Reyes, L. L., & Hernández-Hernández, T. (2015).
- A metacalibrated time-tree documents the early rise of flowering plant phylogenetic
- diversity. New Phytologist, 207(2), 437-453.
- McTavish, E. J., Hinchliff, C. E., Allman, J. F., Brown, J. W., Cranston, K. A., Holder, M.
- T., ... Smith, S. (2015). Phylesystem: A git-based data store for community-curated
- phylogenetic estimates. Bioinformatics, 31 (17), 2794–2800.
- Michonneau, F., Brown, J. W., & Winter, D. J. (2016). rotl: an R package to interact with
- the Open Tree of Life data. Methods in Ecology and Evolution, 7(12), 1476–1481.
- https://doi.org/10.1111/2041-210X.12593
- Morlon, H. (2014). Phylogenetic approaches for studying diversification. Ecology Letters,
- 17(4), 508–525. https://doi.org/10.1111/ele.12251
- Moyle, R. G., Oliveros, C. H., Andersen, M. J., Hosner, P. A., Benz, B. W., Manthey, J. D.,
- 833 ... Faircloth, B. C. (2016). Tectonic collision and uplift of wallacea triggered the

- global songbird radiation. Nature Communications, 7(1), 1–7.
- Oliveros, C. H., Field, D. J., Ksepka, D. T., Barker, F. K., Aleixo, A., Andersen, M. J., ...
- others. (2019). Earth history and the passerine superradiation. Proceedings of the
- National Academy of Sciences, 116(16), 7916–7925.
- Ooms, J., & Chamberlain, S. (2018). *Phylocomr: Interface to 'phylocom'*. Retrieved from https://CRAN.R-project.org/package=phylocomr
- Open Tree Of Life, Redelings, B., Cranston, K. A., Allman, J., Holder, M. T., & McTavish,
- E. J. (2016). Open Tree of Life APIs v3.0. Open Tree of Life Project, (Online
- Resources). Retrieved from
- 843 {https://github.com/OpenTreeOfLife/germinator/wiki/Open-Tree-of-Life-Web-
- 844 APIs}
- Open Tree Of Life, Redelings, B., Sánchez Reyes, L. L., Cranston, K. A., Allman, J., Holder,
- M. T., & McTavish, E. J. (2019). Open tree of life synthetic tree v12.3. Zenodo.
- Retrieved from https://doi.org/10.5281/zenodo.3937742
- Ödeen, A., Håstad, O., & Alström, P. (2011). Evolution of ultraviolet vision in the largest avian radiation-the passerines. *BMC Evolutionary Biology*, 11(1), 1–8.
- Paradis, E., Claude, J., & Strimmer, K. (2004). APE: analyses of phylogenetics and evolution in R language. *Bioinformatics*, 20(2), 289–290.
- Parchman, T. L., Benkman, C. W., & Mezquida, E. T. (2007). Coevolution between
- hispaniolan crossbills and pine: Does more time allow for greater phenotypic
- escalation at lower latitude? Evolution, 61(9), 2142–2153.
- Päckert, M., Martens, J., Sun, Y.-H., Severinghaus, L. L., Nazarenko, A. A., Ting, J., ...
- Tietze, D. T. (2012). Horizontal and elevational phylogeographic patterns of

- himalayan and southeast asian forest passerines (aves: Passeriformes). Journal of
 Biogeography, 39(3), 556–573.
- Pennell, M. W., Eastman, J. M., Slater, G. J., Brown, J. W., Uyeda, J. C., FitzJohn, R. G.,

 1. Harmon, L. J. (2014). Geiger v2. 0: An expanded suite of methods for fitting

 1. macroevolutionary models to phylogenetic trees. *Bioinformatics*, 30 (15), 2216–2218.
- Posadas, P., Crisci, J. V., & Katinas, L. (2006). Historical biogeography: A review of its basic concepts and critical issues. *Journal of Arid Environments*, 66(3), 389–403.
- Powell, A. F., Barker, F. K., Lanyon, S. M., Burns, K. J., Klicka, J., & Lovette, I. J. (2014).

 A comprehensive species-level molecular phylogeny of the new world blackbirds

 (icteridae). *Molecular Phylogenetics and Evolution*, 71, 94–112.
- Powell, C. L. E., Waskin, S., & Battistuzzi, F. U. (2020). Quantifying the error of secondary vs. Distant primary calibrations in a simulated environment. Frontiers in Genetics, 11, 252.
- Price, T. D., Hooper, D. M., Buchanan, C. D., Johansson, U. S., Tietze, D. T., Alström, P.,

 others. (2014). Niche filling slows the diversification of himalayan songbirds.

 Nature, 509 (7499), 222.
- Pulgarín-R, P. C., Smith, B. T., Bryson Jr, R. W., Spellman, G. M., & Klicka, J. (2013).

 Multilocus phylogeny and biogeography of the new world pheucticus grosbeaks (aves:

 Cardinalidae). Molecular Phylogenetics and Evolution, 69(3), 1222–1227.
- Rabosky, D. L. (2015). No substitute for real data: A cautionary note on the use of
 phylogenies from birth–death polytomy resolvers for downstream comparative
 analyses. *Evolution*, 69(12), 3207–3216.
- Rabosky, D. L., Chang, J., Title, P. O., Cowman, P. F., Sallan, L., Friedman, M., ... others.

- (2018). An inverse latitudinal gradient in speciation rate for marine fishes. *Nature*, 559 (7714), 392.
- Ramshaw, J., Richardson, D., Meatyard, B., Brown, R., Richardson, M., Thompson, E., & Boulter, D. (1972). The time of origin of the flowering plants determined by using amino acid sequence data of cytochrome c. New Phytologist, 71(5), 773–779.
- Ratnasingham, S., & Hebert, P. D. (2007). BOLD: The barcode of life data system (http://www. Barcodinglife. Org). *Molecular Ecology Notes*, 7(3), 355–364.
- R Core Team. (2018). R: a language and environment for statistical computing. Vienna,

 Austria: R Foundation for Statistical Computing.
- Rees, & Cranston, K. (2017). Automated assembly of a reference taxonomy for phylogenetic data synthesis. *Biodiversity Data Journal*, (5).
- Rees, Vandepitte, L., Decock, W., & Vanhoorne, B. (2017). IRMNG 2006–2016: 10 Years of a Global Taxonomic Database. *Biodiversity Informatics*, 12.
- Revell, L. J. (2012). Phytools: An r package for phylogenetic comparative biology (and other things). *Methods in Ecology and Evolution*, 3, 217–223.
- Ronquist, F., & Huelsenbeck, J. P. (2003). MrBayes 3: Bayesian phylogenetic inference under mixed models. *Bioinformatics*, 19(12), 1572–1574.

 https://doi.org/10.1093/bioinformatics/btg180
- Roquet, C., Lavergne, S., & Thuiller, W. (2014). One tree to link them all: A phylogenetic dataset for the european tetrapoda. *PLoS Currents*, 6.
- Sanchez-Reyes, L. L., & O'Meara, B. (2022). datelifeplot: Methods to plot chronograms
 and outputs of the datelife package. R Package Release V0.2.2. Retrieved from
 https://zenodo.org/badge/latestdoi/381501451

- Sanchez-Reyes, L. L., O'Meara, B., Eastman, J., Heath, T., Wright, A., Schliep, K., . . .
- Alfaro, M. (2022). datelife: Scientific Data on Time of Lineage Divergence for Your
- Taxa. In R package version 0.6.6. Retrieved from
- https://CRAN.R-project.org/package=datelife and
- 907 https://doi.org/10.5281/zenodo.593938
- Sanderson, M. (2002). Estimating Absolute Rates of Molecular Evolution and Divergence
- Times: A Penalized Likelihood Approach. Molecular Biology and Evolution, 19(1),
- 910 101–109. https://doi.org/10.1093/oxfordjournals.molbev.a003974
- Sanderson, M., & Doyle, J. (2001). Sources of error and confidence intervals in estimating
- the age of angiosperms from rbcL and 18S rDNA data. American Journal of Botany,
- 913 88(8), 1499–1516.
- Sauquet, H. (2013). A practical guide to molecular dating. Comptes Rendus Palevol, 12(6),
- 915 355–367.
- Sauquet, H., Ho, S. Y. W., Gandolfo, M. a, Jordan, G. J., Wilf, P., Cantrill, D. J., ...
- Udovicic, F. (2012). Testing the impact of calibration on molecular divergence times
- using a fossil-rich group: the case of Nothofagus (Fagales). Systematic Biology, 61(2),
- 919 289–313. https://doi.org/10.1093/sysbio/syr116
- Sauquet, H., Ramírez-Barahona, S., & Magallón, S. (2021). The age of flowering plants is
- unknown.
- Schenk, J. J. (2016). Consequences of secondary calibrations on divergence time estimates.
- 923 PLoS ONE, 11(1). https://doi.org/10.1371/journal.pone.0148228
- Schliep, K. P. (2011). Phangorn: Phylogenetic analysis in r. Bioinformatics, 27(4), 592–593.
- Schoch, C. L., Ciufo, S., Domrachev, M., Hotton, C. L., Kannan, S., Khovanskaya, R., . . .

- others. (2020). NCBI Taxonomy: a Comprehensive Update on Curation, Resources and Tools. *Database*, 2020.
- Selvatti, A. P., Gonzaga, L. P., & Moraes Russo, C. A. de. (2015). A paleogene origin for crown passerines and the diversification of the oscines in the new world. *Molecular Phylogenetics and Evolution*, 88, 1–15.
- Shaul, S., & Graur, D. (2002). Playing chicken (gallus gallus): Methodological inconsistencies of molecular divergence date estimates due to secondary calibration points. *Gene*, 300 (1-2), 59–61.
- Smith, S., & Brown, J. (2018). Constructing a broadly inclusive seed plant phylogeny.

 American Journal of Botany, 105(3), 302–314.
- Smith, S., & O'Meara, B. (2012). TreePL: Divergence time estimation using penalized likelihood for large phylogenies. *Bioinformatics*, 28(20), 2689–2690. https://doi.org/10.1093/bioinformatics/bts492
- Stoltzfus, A., Lapp, H., Matasci, N., Deus, H., Sidlauskas, B., Zmasek, C. M., . . . Jordan, G. (2013). Phylotastic! Making tree-of-life knowledge accessible, reusable and convenient.

 BMC Bioinformatics, 14. https://doi.org/10.1186/1471-2105-14-158
- Sun, M., Folk, R. A., Gitzendanner, M. A., Soltis, P. S., Chen, Z., Soltis, D. E., & Guralnick,
 R. P. (2020). Estimating rates and patterns of diversification with incomplete
 sampling: A case study in the rosids. American Journal of Botany, 107(6), 895–909.
- Tietze, D. T., Päckert, M., Martens, J., Lehmann, H., & Sun, Y.-H. (2013). Complete
 phylogeny and historical biogeography of true rosefinches (aves: Carpodacus).

 Zoological Journal of the Linnean Society, 169(1), 215–234.
- Title, P. O., & Rabosky, D. L. (2016). Do Macrophylogenies Yield Stable Macroevolutionary

- Inferences? An Example from Squamate Reptiles. Systematic Biology, syw102. https://doi.org/10.1093/sysbio/syw102
- Treplin, S., Siegert, R., Bleidorn, C., Thompson, H. S., Fotso, R., & Tiedemann, R. (2008).

 Molecular phylogeny of songbirds (aves: Passeriformes) and the relative utility of

 common nuclear marker loci. *Cladistics*, 24(3), 328–349.
- Uyeda, J. C., Pennell, M. W., Miller, E. T., Maia, R., & McClain, C. R. (2017). The evolution of energetic scaling across the vertebrate tree of life. *The American*Naturalist, 190(2), 185–199.
- Vos, R. A., Balhoff, J. P., Caravas, J. A., Holder, M. T., Lapp, H., Maddison, W. P., ...
 others. (2012). NeXML: Rich, extensible, and verifiable representation of
 comparative data and metadata. *Systematic Biology*, 61(4), 675–689.
 https://doi.org/10.1093/sysbio/sys025
- Vos, R. A., & Mooers, A. Ø. (2004). Reconstructing divergence times for supertrees: A

 molecular approach. Phylogenetic Supertrees: Combining Information to Reveal the

 Tree of Life, 281–299.
- Webb, C. (2000). Exploring the Phylogenetic Structure of Ecological Communities: An

 Example for Rain Forest Trees. *The American Naturalist*, 156(2), 145–155.
- Webb, C., Ackerly, D., & Kembel, S. (2008). Phylocom: Software for the analysis of
 phylogenetic community structure and trait evolution. Bioinformatics, 24 (18),
 2098–2100. https://doi.org/10.1093/bioinformatics/btn358
- Webb, C., & Donoghue, M. (2005). Phylomatic: Tree assembly for applied phylogenetics.
 Molecular Ecology Notes, 5(1), 181–183.
- Weir, J., & Schluter, D. (2008). Calibrating the avian molecular clock. *Molecular Ecology*,

976

17(10), 2321-2328.

Zuccon, D., Prŷs-Jones, R., Rasmussen, P. C., & Ericson, P. G. (2012). The phylogenetic
 relationships and generic limits of finches (fringillidae). Molecular Phylogenetics and
 Evolution, 62(2), 581–596.

Figure Captions

Figure 1. DateLife's benchmarking results showing computation time used for taxon 977 name processing and search across datelife's chronogram database, as a function of 978 970 000, we randomly sampled N species names from the class Aves, a hundred times, and 980 then performed a datelife search processing the input names using the Taxon Names 981 Resolution Service (TNRS; light gray), and without processing input names (dark gray). 982 For comparison, we performed a chronogram search using names that have been 983 pre-processed with TNRS. 984

Figure 2. DateLife results of an analysis of a small sample of 6 bird species within the
Passeriformes (a, b). Processed species names were found across 9 chronograms within 6
independent studies (c; Barker et al. (2012), Barker et al. (2015), Burns et al. (2014),
Hedges et al. (2015), Hooper and Price (2017), Jetz et al. (2012). This revealed 28 source
age data points for the queried species names (e; Table 1). Summarized age data (f; Table
2) was used as secondary calibrations to date a tree topology obtained from OpenTree's
synthetic tree v13.4 (d), resulting in the chronogram of summary source ages shown in (g).

Figure 3. Median summary chronogram resulting from a Datelife analysis of bird
species within the family Fringillidae. For visualization purposes, we are showing a portion
of the final median summary chronogram encompassing 57 species out of the 289 total
included in the analysis. The complete final chronogram is available as Supplementary
Figure S3. The starting tree topology (Supplementary Fig. S2) has 289 tip species and 253

nodes; DateLife revealed age data for 194 of these nodes from at least one published 997 chronogram. In total, 19 different chronograms from 13 different studies contributed 818 998 age data points, which were summarized to obtain a single value for each one of the 194 999 nodes with age data. From the 194 summary ages available, 21 were discarded and not 1000 used as calibrations (asterisk, *), because they were older than a parent node or younger 1001 than a descendant node; the remaining 173 summary ages were used as secondary 1002 calibrations (forward slash, /) with the Branch Length Adjuster (BLADJ) software from 1003 Webb et al., (2008). 1004

Figure 4. Cross validation of results from a DateLife analysis of the family

Fringillidae, shown in Figure 3 and Supplementary Figure S3. Each plot compares the

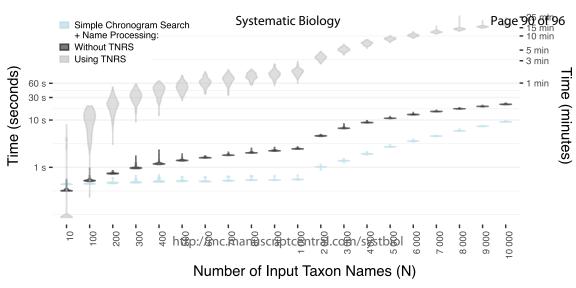
original node age estimates from an input source study chronogram (x axis) with the

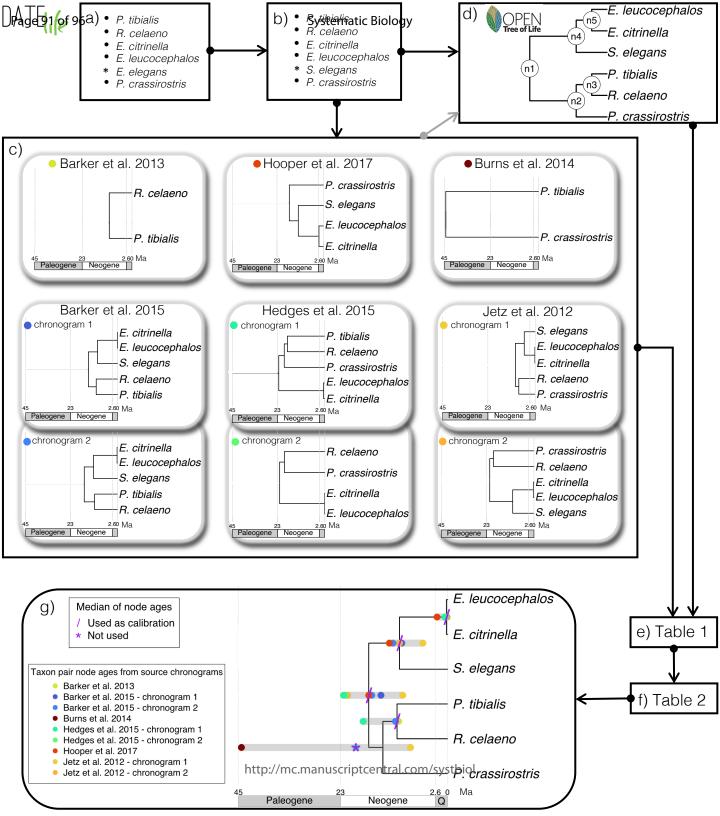
corresponding node age resulting from a dating analysis using the DateLife workflow,

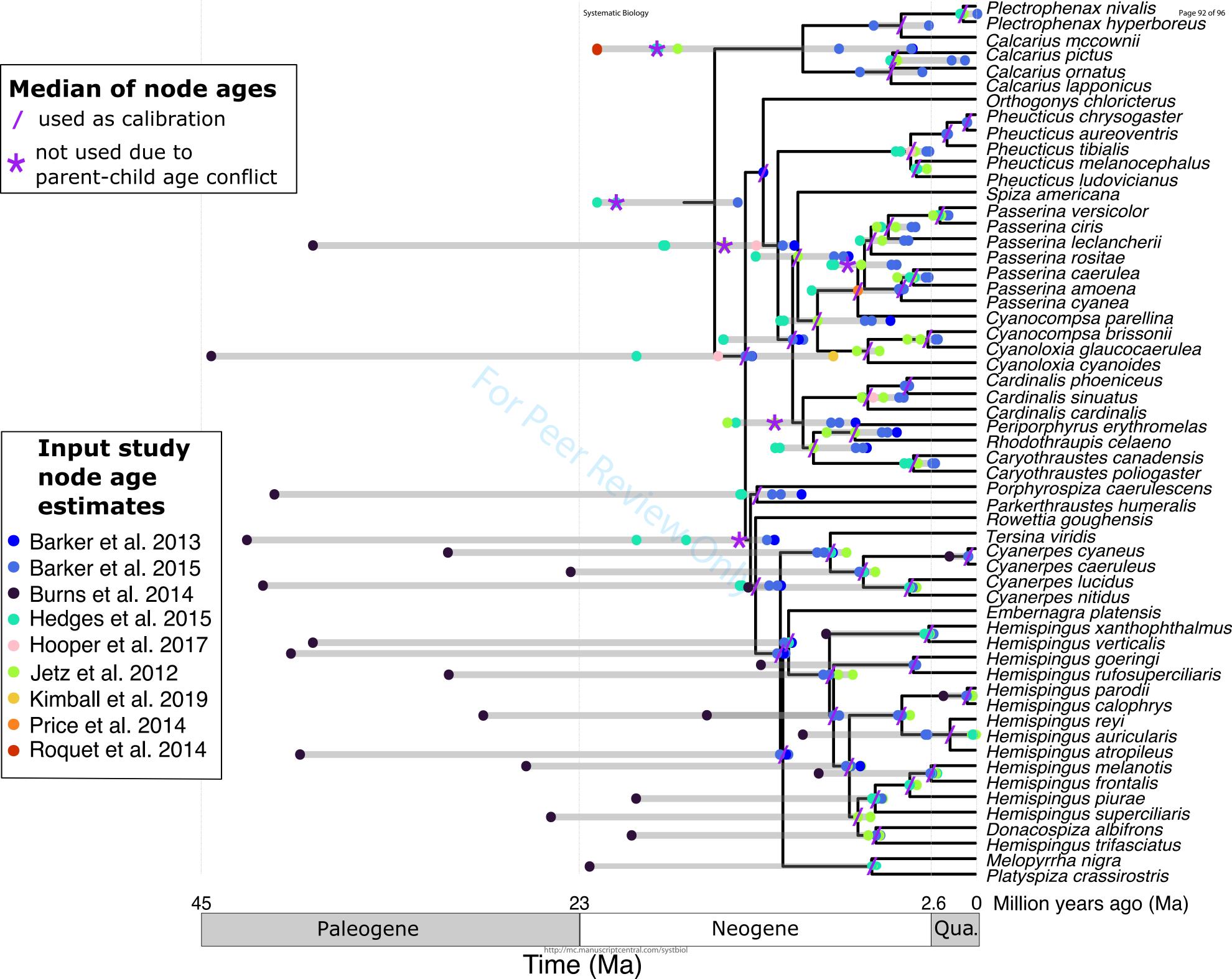
excluding data from that study (y axis).

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









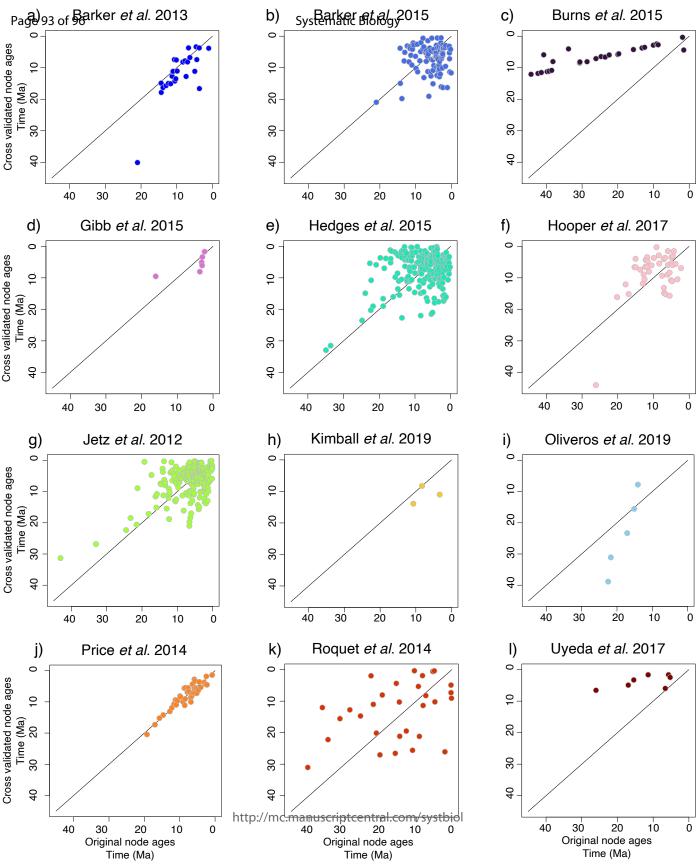


Table 1. Node ages extracted with DateLife from 9 chronograms (shown in Fig. 2c) containing at least 2 of the 6 bird species analysed in the Fringillidae small example. All nodes are congruent with the chosen tree topology shown in Figure 2d.

Node Name	taxon A	taxon B	Node	Study reference
n1	Emberiza citrinella	Platyspiza crassirostris	Age 9.651	Study reference Jetz et al. 2012 - chronogram 1
n1	Pheucticus tibialis	Schoeniclus elegans	14.334	Barker et al. 2015 - chronogram 1
n1	Rhodothraupis celaeno	Schoeniclus elegans	16.298	Barker et al. 2015 - chronogram 2
n1	Emberiza citrinella	Platyspiza crassirostris	16.950	Hooper et al. 2017
n1	Rhodothraupis celaeno	Schoeniclus elegans	21.514	Jetz et al. 2012 - chronogram 2
n1	Emberiza leucocephalos	Platyspiza crassirostris	22.000	Hedges et al. 2015 - chronogram 2
n1	Emberiza citrinella	Platyspiza crassirostris	22.376	Hedges et al. 2015 - chronogram 1
n2	Platyspiza crassirostris	Rhodothraupis celaeno	7.969	Jetz et al. 2012 - chronogram 1
n2	Platyspiza crassirostris	Rhodothraupis celaeno	19.709	Jetz et al. 2012 - chronogram 2
n2	Platyspiza crassirostris	Rhodothraupis celaeno	19.709	Hedges et al. 2015 - chronogram 2
n2	Platyspiza crassirostris	Rhodothraupis celaeno	19.713	Hedges et al. 2015 - chronogram 1
n2	Pheucticus tibialis	Platyspiza crassirostris	44.296	Burns et al. 2014
n3	Pheucticus tibialis	Rhodothraupis celaeno	10.530	Barker et al. 2015 - chronogram 1
n3	Pheucticus tibialis	Rhodothraupis celaeno	10.538	Barker et al. 2013
n3	Pheucticus tibialis	Rhodothraupis celaeno	11.210	Barker et al. 2015 - chronogram 2
n3	Pheucticus tibialis	Rhodothraupis celaeno	18.157	Hedges et al. 2015 - chronogram 1
n4	Emberiza citrinella	Schoeniclus elegans	5.324	Jetz et al. 2012 - chronogram 1
n4	Emberiza	Schoeniclus elegans	9.862	Barker et al. 2015 -

Node			Node					
Name	taxon A	taxon B	Age	Study reference				
	leucocephalos			chronogram 1				
n4	Emberiza leucocephalos	Schoeniclus elegans	10.339	Jetz et al. 2012 - chronogram 2				
n4	Emberiza leucocephalos	Schoeniclus elegans 11.732		Barker et al. 2015 - chronogram 2				
n4	Emberiza citrinella	Schoeniclus elegans	12.513	Hooper et al. 2017				
n5	Emberiza citrinella	Emberiza leucocephalos	0.141	Jetz et al. 2012 - chronogram 1				
n5	Emberiza citrinella	Emberiza leucocephalos	0.152	Hedges et al. 2015 - chronogram 2				
n5	Emberiza citrinella	Emberiza leucocephalos	0.201	Barker et al. 2015 - chronogram 1				
n5	Emberiza citrinella	Emberiza leucocephalos	0.241	Barker et al. 2015 - chronogram 2				
n5	Emberiza citrinella	Emberiza leucocephalos	0.273	Jetz et al. 2012 - chronogram 2				
n5	Emberiza citrinella	Emberiza leucocephalos	0.576	Hedges et al. 2015 - chronogram 1				
n5	Emberiza citrinella	Emberiza leucocephalos	2.290	Hooper et al. 2017				

Table 2. Summary statistics of node ages from Table 1. DateLife uses median ages by default to date a chosen tree topology (Fig. 3g).

Node	Min		Median	Mean		Max		
Name	Age	Q1	Age	Age	Q3	Age	Variance	SD
n1	9.651	15.316	16.950	17.589	21.757	22.376	22.243	4.716
n2	7.969	19.709	19.709	22.279	19.713	44.296	177.328	13.316
n3	10.530	10.536	10.874	12.609	12.946	18.157	13.783	3.713
n4	5.324	9.862	10.339	9.954	11.732	12.513	7.826	2.798
n5	0.141	0.176	0.241	0.553	0.425	2.290	0.608	0.780