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

Abstract

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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 implemented as an R package and an R Shiny web application available at www.datelife.org, that provides services for efficient and easy discovery, 25 summary, reuse, and reanalysis of node age data mined from a curated database of expert, 26 peer-reviewed, and openly available chronograms. The main DateLife workflow starts with 27 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 29 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 to corresponding nodes from a chosen tree topology using the congruification algorithm. Congruified node ages are used as secondary calibrations to date the chosen topology, with or without initial 34 branch lengths, using different phylogenetic dating methods such as BLADJ, treePL, 35 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. Results from any software for scientific analysis can only be as good as the data used as input, we highlight 41 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 increase 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 hypotheses on downstream
- analyses, providing a framework for a more informed interpretation of evolutionary results.
- 47 Keywords: Tree; Phylogeny; Scaling; Dating; Ages; Divergence times; Open Science;
- ⁴⁸ Congruification; Supertree; Calibrations; Secondary calibrations.
- 49 Word count: 6715

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 the
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

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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 three main steps of the DateLife workflow: creating a search query, searching a database, and 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 (TNRS), which increases the probability of correctly finding the queried taxon names in the chronogram database. TNRS detects, corrects and standardizes name misspellings and typos, variant spellings and authorities, and nomenclatural synonyms to a single taxonomic standard (Boyle et al., 2013). TNRS also allows to correctly choose between homonyms, by considering other taxa provided as input to infer the taxonomic context of the 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 taxonomics currently supported by DateLife are the National Center of Biotechnology Information (NCBI) taxonomic database (Schoch et al., 2020), the Global Biodiversity Information Facility (GBIF) taxonomic backbone (GBIF Secretariat, 2022), and the Interim

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

Besides binomial species names, DateLife accepts scientific names from any inclusive 102 taxonomic group (e.g., genus, family, tribe), as well as subspecific taxonomic variants (e.g., 103 subspecies, variants, strains). If a taxon name belongs to an inclusive taxonomic group, 104 DateLife has two alternative behaviors defined by the "get species from taxon" flag. If the 105 flag is active, DateLife retrieves all species names within a taxonomic group provided, from a 106 standard taxonomy of choice, and adds them to the search query. In this case, subspecific 107 variants are excluded. If the flag is inactive, DateLife excludes inclusive taxon names from 108 the search query, and species and subspecific variant names are processed as provided by the 100 user. The processed taxon names are saved as an R object of a newly defined class, 110 datelifeQuery, that is used in the following steps. This object contains the input names 111 standardized to a taxonomy of choice (OTT by default), the corresponding OTT id numbers, 112 and the topology of an input tree, if one was provided. 113

Searching a Chronogram Database

At the time of writing of this manuscript (May 15, 2023), DateLife's chronogram 115 database latest version consist of 253 chronograms published in 187 different studies. It is 116 curated from OpenTree's phylogenetic database, the Phylesystem, which constitutes an open source of expert and peer-reviewed phylogenetic knowledge with rich metadata (McTavish et al., 2015), which allows automatic and reproducible assembly of our chronogram database. 119 Datelife's chronogram database is navigable as an R data object within the datelife R package.

A unique feature of the Phylesystem is that any user can add new published, 122 state-of-the-art chronograms any time, through OpenTree's curator application 123 (https://tree.opentreeoflife.org/curator). As chronograms are added to Phylesystem, they 124 can be incorporated into the chronogram database of the datelife R package, which is 125

currently manually updated as new chronogram data is added to Phylesystem. The updated
database is assigned a new version number, followed by a package release on CRAN. Users
can directly run datelife functions to trigger an update of their local chronogram database,
to incorporate any new chronograms to their DateLife analysis before an official database
update is released on CRAN.

A DateLife search is implemented by matching processed taxon names provided by the 131 user to tip labels in the chronogram database. Chronograms with at least two matching 132 taxon names on their tip labels are identified and pruned down to preserve only the matched 133 taxa. These matching pruned chronograms are referred to as source chronograms. Total 134 distance in units of million years (Myr) between taxon pairs within each source chronogram 135 are stored as a patristic distance matrix (Fig. 1). The matrix format speeds up extraction of 136 pairwise taxon ages of any queried taxa, as opposed to searching the ancestor node of a pair 137 of taxa in a "phylo" object or newick string. Finally, the patristic matrices are associated to 138 the study citation where the original chronogram was published, and stored as an R object of the newly defined class datelifeResult. 140

Summarizing Search Results

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

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

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- 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 156 users provide as input a tree topology from the literature, or one of their own making. If no 157 topology is provided, DateLife automatically extracts one from the OpenTree synthetic tree, 158 a phylogeny currently encompassing 2.3 million taxa across all life, assembled from 1, 239 159 published phylogenetic trees and OpenTree's unified Taxonomy, OTT (Open Tree Of Life et 160 al., 2019). Alternatively, DateLife can combine topologies from source chronograms using a 161 supertree approach (Criscuolo, Berry, Douzery, & Gascuel, 2006). To do this, DateLife first 162 identifies the source chronograms that form a grove, roughly, a sufficiently overlapping set of 163 taxa between trees, by implementing definition 2.8 for n-overlap from Ané et al. (2009). If 164 the source chronograms do not form a grove, the supertree reconstruction will fail. In rare 165 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 supertree) topology, that combines topologies from source chronograms in a grove. 169

Applying Secondary Calibrations

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

present. Hence, node ages correspond to half the values stored in the datelifeResult matrices. A table of congruified node ages that can be used as calibrations for a dating analysis is stored as a congruifiedCalibrations object.

For each congruent node, the pairwise distances that traverse that node are summarized 179 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), 181 which deforms patristic distance matrices by minimizing variance and then averaging them. 182 These single summary taxon pair age matrices are stored as summarized calibrations that 183 can be used as secondary calibrations to date a tree topology - with or without initial branch 184 lengths, using phylogenetic dating methods currently supported within DateLife: BLADJ 185 (Webb, Ackerly, & Kembel, 2008; Webb & Donoghue, 2005), MrBayes (Huelsenbeck & 186 Ronquist, 2001; Ronquist & Huelsenbeck, 2003), PATHd8 (Britton, Anderson, Jacquet, 187 Lundqvist, & Bremer, 2007), and treePL (Smith & O'Meara, 2012). 188

Dating a Tree Topology

Dating a tree with no branch lengths.—When producing or obtaining a tree with 190 branch lengths for a group of interest is not possible. DateLife can date a topology without 191 branch lengths, obtained from OpenTree or by implementing the supertree approach 192 described above, by applying the Branch Length Adjuster (BLADJ; Webb et al., 2008; Webb 193 & Donoghue, 2005) algorithm, which requires no initial branch lengths. The algorithm starts 194 by fixing ages for nodes with calibration data upon the given topology. Then, it distributes 195 time for nodes with no data evenly between calibrated nodes, minimizing age variance in the resulting chronogram. This approach has proven useful for ecological analyses that require a phylogenetic time context (Webb et al., 2008). When there is conflict between ages of 198 calibrated nodes, BLADJ ignores node ages that are older than the age of a parent node. 199 The BLADJ algorithm requires a root age estimate to run. If there is no information on the 200 age of the root in the chronogram database, users can provide an estimate from the 201

literature. If none is provided, DateLife will not return a dated topology and will provide a warning message along with suggestions on how the user can provide an age for the root so that a fully dated tree can be returned.

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 common practice in the literature to use a birth-death model to assign ages to nodes with no genetic data (Jetz, Thomas, Joy, Hartmann, & Mooers, 2012; Rabosky et al., 2018; Smith & Brown, 2018). To do so, DateLife implements MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), using nodes with published age data as calibration priors on a fixed topology, a simple birth-death model with parameters that can be determined by the user, and no genetic data.

Dating a tree with branch lengths. - Relative branch lengths can provide key 213 information for phylogenetic dating, specifically for nodes without any calibration data 214 available. While using initial branch length data is the golden standard for phylogenetic 215 dating analyses, estimating trees with branch lengths proportional to substitution rates per 216 site requires obtaining primary data, assembling and curating a homology hypothesis, and 217 choosing and implementing a method for phylogenetic inference. DateLife implements a 218 workflow to streamline this process by applying open data from the Barcode of Life Data 219 System, BOLD (Ratnasingham & Hebert, 2007) to obtain genetic markers for input taxa. 220 By default, BOLD genetic sequences are aligned with MUSCLE (Edgar, 2004) using 221 functions from the msa R package (Bodenhofer et al., 2015). Alternatively, sequences can be aligned with MAFFT (Katoh, Asimenos, & Toh, 2009), using functions from the ape R 223 package (Paradis et al., 2004). The BOLD sequence alignment is then used to obtain initial branch lengths with the accelerated transformation (ACCTRAN) parsimony algorithm, 225 which resolves ambiguous character optimization by assigning changes along branches of the 226 tree as close to the root as possible (Agnarsson & Miller, 2008), resulting in older internal 227

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 220 topology and a simple Jukes-Cantor model, producing a BOLD tree with branch lengths 230 proportional to expected number of substitutions per site. Both parsimony and ML 231 optimizations are done with functions from the phangorn package (Schliep, 2011). Due to 232 the computing load it requires, the BOLD workflow is currently only supported through 233 DateLife's R package. It is not yet available through the web application. 234

Phylogenetic dating methods supported in DateLife that incorporate branch length 235 information from the input topology in combination with the secondary calibrations include: 236 PATHd8, a non-clock, rate-smoothing method to date trees (Britton et al., 2007); treePL 237 (Smith & O'Meara, 2012), a semi-parametric, rate-smoothing, penalized likelihood dating 238 method (Sanderson, 2002); and MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & 239 Huelsenbeck, 2003), a Bayesian inference program implementing Markov chain Monte Carlo 240 (MCMC) methods to estimate a posterior distribution of model parameters. 241

Visualizing Results

Finally, users can save all source and summary chronograms in formats allowing for 243 reuse and reanalysis, such as newick and the R "phylo" format. Input and summary 244 chronograms can be visualized and compared graphically, and users can construct their own 245 graphs using DateLife's chronogram plot generation functions available from the R package 246 datelifeplot (Sanchez-Reyes & O'Meara, 2022).

BENCHMARK 248

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R package datelife code speed was tested on an Apple iMac with one 3.4 GHz Intel 249 Core is processor. We registered variation in computing time of query processing and search 250 through the database relative to number of queried taxon names. Query processing time 251 increases roughly linearly with number of input taxon names, and increases considerably if 252

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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. 2).

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 3. As a second example, we analysed 289 bird species in the family Fringillidae that are included in the NCBI taxonomy. The resulting summary chronogram is shown in Figure 4, and results from previous steps of the workflow are available as Supplementary Figures.

A Small Example

Creating a search query.— We chose 6 bird species within the Passeriformes. The
sample includes two species of cardinals: the black-thighed grosbeak — Pheucticus tibialis
and the crimson-collared grosbeak — Rhodothraupis celaeno; three species of buntings: the
yellowhammer — Emberiza citrinella, the pine bunting — Emberiza leucocephalos and the
yellow-throated bunting — Emberiza elegans; and one species of tanager, the vegetarian finch —

Platyspiza crassirostris. Processing of input names found that Emberiza elegans is synonym for Schoeniclus elegans in the default reference taxonomy (OTT v3.3, June 1, 2021). For a detailed discussion on the state of the synonym, refer to Avibase (Avibase, 2022; Lepage, 2004; Lepage, Vaidya, & Guralnick, 2014). Discovering this synonym allowed assigning five age data points for the parent node of Emberiza elegans, shown as Schoeniclus elegans in Figure 3A, which would not have had any data otherwise.

Searching the database. DateLife used the processed input names to search the 284 local chronogram database and found 9 matching chronograms from 6 different studies (Fig. 285 3B). Three studies matched five input names (Barker, Burns, Klicka, Lanyon, & Lovette, 286 2015; Hedges, Marin, Suleski, Paymer, & Kumar, 2015; Jetz et al., 2012), one study matched 287 four input names (Hooper & Price, 2017) and two studies matched two input names (Barker, 288 Burns, Klicka, Lanyon, & Lovette, 2013; Burns et al., 2014). No studies matched all input 289 names. Together, source chronograms provide 28 unique age data points, covering all nodes 290 on our chosen tree topology to date (Table 1). 291

Summarizing search results. - DateLife obtained OpenTree's synthetic tree 292 topology for these taxa (Fig. 3C), and congruified and mapped age data to nodes in this 293 chosen topology (Table 1). The name processing step allowed including five data points for 294 node "n4" (parent of Schoeniclus elegans; Fig. 3A) that would not have had any data 295 otherwise due to name mismatch. Age summary statistics per node were calculated (Table 2) 296 and used as calibrations to date the tree topology using the BLADJ algorithm. As expected, 297 more inclusive nodes (e.g., node "n1") have more variance in age data than less inclusive 298 nodes (e.g., node "n5"). Summary age data for node "n2" were excluded as final calibration 299 because they are older than age data of the more inclusive node, "n1" (Fig. 3C4).

An Example with the Family of True Finches

Creating a query. To obtain ages for all species within the family of true finches, 302 Fringillidae, we ran a DateLife query using the "get species from taxon" flag, which gets all 303 recognized species names within a named group from a taxonomy of choice. Following the 304 NCBI taxonomy, our DateLife query has 289 Fringillidae species names. This 305 taxon-constrained approach implies that the full DateLife analysis will be performed using a 306 tree topology and ages available for species names from a given taxonomic group, which do 307 not necessarily correspond to a monophyletic group. Users can change this behavior by 308 providing all species names corresponding to a monophyletic group as input for a DateLife 300 search, or a monophyletic tree to construct a DateLife summary. 310

Searching the database. Next, we used the processed species names in our 311 DateLife query to identify chronograms with at least two Fringillidae species as tip taxa. 312 The DateLife search identified 19 chronograms matching this criteria, published in 13 313 different studies (Barker et al., 2013, 2015; Burns et al., 2014; Claramunt & Cracraft, 2015; 314 Gibb et al., 2015; Hedges et al., 2015; Hooper & Price, 2017; Jetz et al., 2012; Kimball et al., 315 2019; Oliveros et al., 2019; Price et al., 2014; Roquet, Lavergne, & Thuiller, 2014; Uyeda, 316 Pennell, Miller, Maia, & McClain, 2017). Once identified, DateLife pruned these matching 317 chronograms to remove tips that do not belong to the queried taxon names, and transformed 318 these pruned chronograms to pairwise distance matrices, revealing 1, 206 different age data 319 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. 5). Hence, a topology containing only the 289 species from the original query was extracted from Open Tree of Life's synthetic tree v12.3 (Open Tree Of Life et al., 2019).

Age data from source chronograms was congruified to OpenTree's topology (Fig. 5B), reducing the age data set to 818 different data points (Supplementary Table S2). For each congruent node, age summary statistics were calculated and used as fixed secondary calibrations over the chosen tree topology, to obtain a fully dated phylogeny with the program BLADJ (Fig. 4).

Cross-Validation Test

We performed a cross validation test of a DateLife analysis using the Fringillidae 336 source chronograms obtained above (Supplementary Figs. S2-S20). We used as inputs for a 337 DateLife analysis all individual tree topologies from each of the 19 source chronograms from 338 13 studies, treating their node ages as unknown. We congruified node ages extracted from 339 chronograms from all other studies upon the individual topologies, effectively excluding 340 original ages from each topology. Finally, average node ages per node were applied as 341 secondary calibrations and smoothed with the BLADJ algorithm. We found that node ages 342 from the original studies, and ages estimated using all other age data available are largely 343 correlated (Fig. 5). 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 345 original estimate, and overestimate ages for nodes closer to the tips. Accordingly, root ages are generally older in the original study than estimated using cross-validated ages (Supplementary Fig. S1). In general, topologically deeper nodes display the largest age 348 variation between node ages from the original chronograms and ages summarized with 349 DateLife. 350

DISCUSSION

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

A total of 99,474 unique terminal taxa are represented in DateLife's database. 358 Incorporation of more chronograms into the database will continue to improve DateLife's 359 services. One option to increase the number of chronograms in the DateLife database is the Dryad data repository. Methods to automatically mine chronograms from Dryad could be designed and implemented. However, Dryad's metadata system has no information to 362 automatically detect branch length units, and those would still need to be determined 363 manually by a human curator. We would like to emphasize on the importance of sharing 364 chronogram data, including systematically curated metadata, into open repositories, such as 365 OpenTree's Phylesystem (McTavish et al., 2015) for the benefit of the scientific community 366 as a whole. 367

As we envision that DateLife will have many interesting applications in research and 368 beyond, we emphasize that DateLife's results –as well as any insights gleaned from them, 369 largely depend on the quality of the source chronograms: low quality chronograms will 370 produce low quality results. This "garbage in, garbage out" problem has long been 371 recognised in supertree methods for summarizing phylogenetic trees (Bininda-Emonds et al., 2004). We note that this is a surfacing issue of any automated tool for biological data 373 analysis. For example, DNA riddled with sequencing errors will produce generally poor alignments that will return biased evolutionary hypothesis, independently of the quality of the analysis software used. Again, we urge readers and DateLife users to explore all input 376 chronograms before using a summary chronogram resulting from a DateLife workflow. 377

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

Age Variation in Source Chronograms

Conflict in estimated ages among alternative studies is common in the literature. See, 386 for example, the robust ongoing debate about crown group age of angiosperms 387 (Barba-Montoya, Reis, Schneider, Donoghue, & Yang, 2018; Magallón, Gómez-Acevedo, 388 Sánchez-Reyes, & Hernández-Hernández, 2015; Ramshaw et al., 1972; Sanderson & Doyle, 389 2001; Sauquet, Ramírez-Barahona, & Magallón, 2021). Alternative source chronograms 390 available for the same taxa have potentially been estimated implementing different types of 391 calibrations, which affects the resulting node age estimates. For example, in the DateLife 392 analysis of the Fringillidae shown above, the chronograms from one study (Burns et al., 2014) 393 were inferred using molecular substitution rate estimates across birds (Weir & Schluter, 394 2008), and have much older age estimates for the same nodes than chronograms that were 395 inferred using fossil calibrations (Figs. 4, 5; Supplementary Figs. S1, S5). Another source of 396 conflict in estimated node ages can arise from different placements for the same calibration, 397 which would imply fundamentally distinct evolutionary hypotheses (Antonelli et al., 2017). For example, two independent researchers working on the same clade should both carefully select and justify their choices of fossil calibration placement. Yet, if one researcher concludes that a fossil should calibrate the ingroup of a clade, while another researcher concludes that 401 the same fossil should calibrate the outgroup of the clade, the resulting age estimates will 402 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
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
without calibrations but does not necessarily improve accuracy of the estimated ages.

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

Primary vs Secondary Calibrations

While most chronograms in DateLife's database are constructed using primary 416 calibrations, i.e., molecular substitution rates or absolute ages in Myr obtained from the 417 fossil record or geological events, DateLife also constructs summary chronograms using 418 secondary calibrations, i.e., node ages in Myr extracted from existing chronograms. The 419 scientific community has generally more confidence in chronograms generated from a single 420 analysis, where carefully chosen primary calibrations constitute the main source of absolute 421 time information used in the dating analysis (Antonelli et al., 2017; Garzón-Orduña, 422 Silva-Brandão, Willmott, Freitas, & Brower, 2015; Graur & Martin, 2004; Sauquet, 2013; 423 Sauguet et al., 2012; Schenk, 2016; Shaul & Graur, 2002). However, implementation of primary calibrations does not come without difficulties: it requires high expertise and training to discover, place and apply calibrations appropriately (Hipsley & Müller, 2014; Ksepka et al., 2011). One way around it is to use fossils that have been widely discussed and 427 previously curated as calibrations to date other trees (Ksepka et al., 2011; Sauquet, 2013), 428 and making sure that all data reflect a coherent evolutionary history (Sauquet, 2013), as for 429

example done by Antonelli et al. (2017). The Fossil Calibration Database provides 430 availability of 220 primary calibration points encompassing flowering plants and metazoans, 431 that have been curated by experts and used for dating analysis in peer-reviewd publications 432 (Ksepka et al., 2015). This database facilitates the use of expert primary fossil calibrations in 433 new phylogenetic dating analyses. Yet, users still require the expertise to locate appropriate 434 nodes in their phylogenies to corresponding fossils available in the database, potentially 435 delaying increased usage of the database in research. Open databases for effective reuse of 436 fossil calibration data still have room to grow and should constitute a key addition to the 437 phylogenetics tool set. 438

In some cases, secondary calibrations might represent the only information available for 439 a dating analysis (e.g., groups with a poor fossil record). Recently, Powell, Waskin, and Battistuzzi (2020) showed in a simulation analysis that secondary calibrations are as good as 441 primary calibrations. It seems that using several secondary calibrations (as opposed to just 442 one) can provide sufficient information to alleviate or even neutralize potential biases described before (Graur & Martin, 2004; Sauquet, 2013; Shaul & Graur, 2002). Our cross 444 validation analysis also provides some insight into the application of secondary calibrations. 445 In general, node ages summarized with DateLife and those from the original studies appear 446 positively correlated (Supplementary Figs. S2-S20). We also note that nodes closer to the 447 root tend to be slightly younger than ages from the original studies. In contrast, nodes closer 448 to the tips tend to be slightly older when estimated using secondary calibrations than ages 449 from the original studies. The only exception to this trend was observed in Burns et al. 450 (2014) chronogram, which generally displays much younger node ages when estimated using 451 secondary calibrations than the original study (Supplementary Figs. S1, S5), supporting 452 previous observations (Sauguet et al., 2012; Schenk, 2016). 453

We unite to the call of our colleagues: further research is needed to fully understand the effects of using secondary calibrations with alternative tree dating methods, as well as

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usage of resulting chronograms in downstream analyses (Hipsley & Müller, 2014; Powell et al., 2020; Schenk, 2016; Shaul & Graur, 2002). It is possible that secondary calibrations can be used successfully with dating methods that do not require setting priors, such as penalized likelihood (Sanderson, 2003); with methods that do not make any assumptions on the ages and fix them to a node on a tree topology, such as BLADJ (Webb et al., 2008; Webb & Donoghue, 2005); and with methods that summarize age data unto a tree topology, such as DateLife.

Sumarizing Chronograms

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

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

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

DateLife provides an opportunity to capture concordance and conflict among date estimates, which can also be used as a metric for chronogram reliability. Its open database of chronograms allows other researchers to do such analyses themselves reproducibly, and

without needing permission. Though, of course, they should follow proper citation practices,
especially for the source chronogram studies.

The exercise of summarizing age data from across multiple studies provides the 483 opportunity to work with a chronogram that reflects a unified evolutionary history for a 484 lineage, by putting together evidence from different hypotheses. The largest, and 485 taxonomically broadest chronogram currently available from OpenTree was constructed 486 summarizing age data from 2,274 published chronograms using NCBI's taxonomic tree as 487 backbone (Hedges et al., 2015). A summarizing exercise may also amplify the effect of 488 uncertainty and errors in source data, and blur parts of the evolutionary history of a lineage 480 that might only be reflected in source chronograms and lost on the summary chronogram 490 (Sauguet et al., 2021). 491

Effects of Taxon Sampling on Downstream Analyses

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

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

506 353, 185 seed plants where only 23% had molecular data. These stochastically resolved
507 chronograms return diversification rates estimates that appear less biased than those
608 estimated from their incompletely sampled counterparts, even with methods that account for
609 missing lineages by using sampling fractions (Chang et al., 2020; Cusimano et al., 2012), but
610 can also introduce spurious patterns of early bursts of diversification (Cusimano & Renner,
611 2010; Sun et al., 2020).

Taxonomy-based stochastic polytomy resolvers also introduce topological differences in 512 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 515 of a biological group. For example, alternative topologies in chronograms from the same 516 biological group can infer very different species diversification patterns (Rabosky, 2015; Title 517 & Rabosky, 2016). Similarly, there are worries that patterns of morphological evolution 518 cannot be accurately inferred with phylogenies that have been resolved stochastically over a 519 taxonomic backbone, as any patterns would be erased by randomization (Rabosky, 2015). 520 We note that the same applies for geography- and morphology-dependent diversification 521 analysis. Hence, we suggest that phylogenies that have been processed with taxonomy-based 522 stochastic polytomy resolvers, including certain summary chronograms from a DateLife 523 analysis, can be useful as null or neutral models, representing the case of a diversification 524 process that is independent of traits and geographical scenario. 525

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

553 AVAILABILITY

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

repository (Sanchez-Reyes et al., 2022). DateLife's web application is maintained using
RStudio's shiny server and the shiny package open infrastructure, as well as Docker and
OpenTree's infrastructure (datelife.opentreeoflife.org). datelife's stable version can be
installed from the CRAN repository using the command install.packages(pkgs =
"datelife") from within R. Development versions are available from DateLife's GitHub
repository (https://github.com/phylotastic/datelife) and can be installed using the
command devtools::install_github("phylotastic/datelife").

SUPPLEMENTARY MATERIAL

Supplementary Figures can be viewed and dowloaded from their Zenodo repository 565 (https://doi.org/10.5281/zenodo.6683667). Supplementary material, including code, 566 biological examples, benchmark results, data files and online-only appendices, can be 567 downloaded from the Dryad data repository (https://doi.org/10.5061/dryad.cnp5hqc6w), as 568 well as in the Zenodo stable repositories that host the reproducible manuscript 569 (https://doi.org/10.5281/zenodo.7435094), the biological examples 570 (https://doi.org/10.5281/zenodo.7435101), and the software benchmark 571 (https://doi.org/10.5281/zenodo.7435106). Development versions corresponding to all of the 572 above are hosted on GitHub, accesible at https://github.com/LunaSare/datelifeMS1, 573 https://github.com/LunaSare/datelife examples, and 574 https://github.com/LunaSare/datelife_benchmark. 575

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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.
- 607 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
- 610 {https://evolution.genetics.washington.edu/phylip/newicktree.html}
- Avibase. (2022). Yellow-throated Bunting. Avibase the World Bird Database, (Online
- Resource). Retrieved from
- 613 {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).
- 618 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).
- 643 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),
- 646 741–752. 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.
- ⁶⁵⁰ 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.
- ⁶⁵⁶ Chamberlain, S. (2018). bold: Interface to Bold Systems API. Retrieved from
- https://CRAN.R-project.org/package=bold
- 658 Chamberlain, S. A., & Szöcs, E. (2013). taxize: taxonomic search and retrieval in R [version
- 2; referees: 3 approved]. F1000Research, 2(191), 1–29.
- 660 https://doi.org/10.12688/f1000research.2-191.v2
- 661 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.
- 664 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., Marin, J., Suleski, M., Paymer, M., & Kumar, S. (2015). Tree of life reveals clock-like speciation and diversification. *Molecular Biology and Evolution*, 32(4),

- 711 835–845. 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.
- 721 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.
- 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.
- 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.,

 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
- 783 {https://github.com/OpenTreeOfLife/germinator/wiki/Open-Tree-of-Life-Web-
- 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),

- 850 101–109. https://doi.org/10.1093/oxfordjournals.molbev.a003974
- Sanderson, M. (2003). r8s: Inferring Absolute Rates of Molecular Evolution and Divergence
 Times in the Absence of a Molecular Clock. *Bioinformatics*, 19(2), 301–302.
- 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*,
 88(8), 1499–1516.
- Sauquet, H. (2013). A practical guide to molecular dating. Comptes Rendus Palevol, 12(6), 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.

 **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*,

 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.