- 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 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,

- 43 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
- 45 among alternative hypotheses of dates for the same diversification events, and to support
- exploration of the effect of alternative chronogram hypotheses on downstream analyses,
- 47 providing a framework for a more informed interpretation of evolutionary results.
- 48 Keywords: Tree; Phylogeny; Scaling; Dating; Ages; Divergence times; Open Science;
- Congruification; Supertree; Calibrations; Secondary calibrations.
- 50 Word count: 7530

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

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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). 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 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 (Mar 08, 2024), DateLife's chronogram 116 database latest version consist of 253 chronograms published in 187 studies, encompassing 99 117 474 species. It is curated from OpenTree's phylogenetic database, the Phylogystem, an open 118 database of expert and peer-reviewed phylogenetic knowledge with rich metadata and a wide 119 taxonomic scope (McTavish et al., 2015). We expect DateLife's database to largely overlap 120 with OpenTree's phylogenetic database taxonomic coverage, where Chordata and Embryophyta are nearly fully sampled. In contrast, Bacteria, Fungi, Nematoda, and Insecta, 122 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 124 the former groups will get results from a DateLife analysis. If none of the user's species are 125 found, the software will indicate the lack of age data for the queried taxa in the database.

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A unique feature of the Phylesystem is that any user can add new published, 127 state-of-the-art chronograms any time, through OpenTree's curator application 128 (https://tree.opentreeoflife.org/curator). Relying on an open source database permits an 129 automatic and reproducible assembly of DateLife's chronogram database, which is stored and 130 navigable as an R data object within the datelife R package. As chronograms are added to 131 Phylesystem, they can be incorporated into the chronogram database within the datelife 132 R package, by manually triggering an update. The updated datelife database is assigned a 133 new version number, followed by a package release on CRAN. We encourage users to submit 134 published chronograms to OpenTree's phylogenetic database, so that their taxon of interest 135 can be included in future DateLife searches. Users can directly run datelife functions to 136 trigger an update of their local chronogram database, to incorporate any new chronograms 137 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 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 in a "phylo" object or newick string. Finally, the patristic matrices are associated to the study citation where the original chronogram was published, and stored as an R object of the newly defined class datelifeResult.

#### Summarizing Search Results

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

- 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 164 users provide as input a tree topology from the literature, or one of their own making. If no 165 topology is provided, DateLife automatically extracts one from the OpenTree synthetic tree, 166 a phylogeny currently encompassing 2.3 million taxa across all life, assembled from 1,239 167 published phylogenetic trees and OpenTree's unified Taxonomy, OTT (Open Tree Of Life et 168 al., 2019). Alternatively, DateLife can combine topologies from source chronograms using a supertree approach (Criscuolo, Berry, Douzery, & Gascuel, 2006). To do this, DateLife first 170 identifies the source chronograms that form a grove, roughly, a sufficiently overlapping set of 171 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 173 cases, a group of trees can have multiple groves. By default, DateLife chooses the grove with 174 the most taxa, however, the "criterion = trees" flag allows the user to choose the grove with 175 the most trees instead. The result is a single summary (i.e., supertree) topology, that 176 combines topologies from source chronograms in a grove. 177

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Once a topology is chosen, DateLife applies the congruification method (Eastman, 179 Harmon, & Tank, 2013) that find nodes belonging to the same clade across source 180 chronograms, and then extracts the corresponding node ages from patristic distance matrices 181 stored as a datelifeResult object. Note that by definition, these matrices store total 182 distance (time from tip to tip), assuming that the terminal taxa are coeval and occur at the 183 present. Hence, node ages correspond to half the values stored in the datelifeResult 184 matrices. A table of congruified node ages that can be used as calibrations for a dating 185 analysis is stored as a congruifiedCalibrations object. 186

For each congruent node, the pairwise distances that traverse that node are summarized 187 into a single summary matrix using classic summary statistics (i.e., mean, median, minimum 188 and maximum ages), and the Supermatrix Distance Method (SDM; Criscuolo et al., 2006), 189 which deforms patristic distance matrices by minimizing variance and then averaging them. 190 These single summary taxon pair age matrices are stored as summarized calibrations that 191 can be used as secondary calibrations to date a tree topology - with or without initial branch 192 lengths, using phylogenetic dating methods currently supported within DateLife: BLADJ 193 (Webb, Ackerly, & Kembel, 2008; Webb & Donoghue, 2005), MrBayes (Huelsenbeck & Ronquist, 2001; Ronquist & Huelsenbeck, 2003), PATHd8 (Britton, Anderson, Jacquet, 195 Lundqvist, & Bremer, 2007), and treePL (Smith & O'Meara, 2012).

# Dating a Tree Topology

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Dating a tree without branch lengths.— To date a tree topology when initial
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 205 analyses that require a phylogenetic time context (Webb et al., 2008). When there is conflict 206 between ages of calibrated nodes, BLADJ ignores node ages that are older than the age of a 207 parent node. The BLADJ algorithm requires a root age to run. Users can provide an 208 appropriate root age estimate of their own or one obtained from the literature. If a root age 200 is not provided and there is no information on the age of the root in the chronogram 210 database, DateLife chooses a random age for the root, so that a dated tree topology can be 211 generated with BLADJ. In this case, DateLife will provide a conspicuous warning message, 212 so that users are aware that the root of the chronogram was chosen at random because there 213 was no information available for it in the chronogram database, along with suggestions on 214 how the user can find and provide an appropriate age for the root of the initial topology. 215

An alternative to BLADJ to date tree topologies in the absence of initial branch 216 lengths that is common practice in the literature is to use a birth-death model to draw 217 branch lengths (Jetz, Thomas, Joy, Hartmann, & Mooers, 2012; Rabosky et al., 2018; Smith 218 & Brown, 2018). In addition to the initial tree topology and nodes with age data, these 219 methods require initial values of speciation and extinction rate parameters provided by the 220 user. DateLife implements this approach with MrBayes (Huelsenbeck & Ronquist, 2001; 221 Ronquist & Huelsenbeck, 2003), using nodes with published age data as calibration priors on 222 nodes of a tree topology with no branch lengths, a simple birth-death model with speciation 223 and extinction rate parameters that are provided by the user, and no genetic data. However, 224 BLADJ is the default option in DateLife, as it does not require any information on 225 diversification rates for the phylogenetic sample to draw from a branch length distribution. 226

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 231 choosing and implementing a method for phylogenetic inference. DateLife implements a 232 workflow to streamline this process by applying open data from the Barcode of Life Data 233 System, BOLD (Ratnasingham & Hebert, 2007) to obtain genetic markers for input taxa. 234 By default, BOLD genetic sequences are aligned with MUSCLE (Edgar, 2004) using 235 functions from the msa R package (Bodenhofer et al., 2015). Alternatively, sequences can be 236 aligned with MAFFT (Katoh, Asimenos, & Toh, 2009), using functions from the ape R 237 package (Paradis et al., 2004). The BOLD sequence alignment is then used to obtain initial 238 branch lengths with the accelerated transformation (ACCTRAN) parsimony algorithm, 230 which resolves ambiguous character optimization by assigning changes along branches of the 240 tree as close to the root as possible (Agnarsson & Miller, 2008), resulting in older internal 241 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 243 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 245 optimizations are done with functions from the phangorn package (Schliep, 2011). Due to 246 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.

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).

262 Benchmark

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R package datelife code speed was tested on an Apple iMac with one 3.4 GHz Intel 263 Core is processor. We registered variation in computing time of query processing and search 264 through the database relative to number of queried taxon names. Query processing time 265 increases roughly linearly with number of input taxon names, and increases considerably if 266 Taxonomic Name Resolution Service (TNRS) is activated. Up to ten thousand names can be 267 processed and searched in less than 30 minutes with the most time consuming settings. Once 268 names have been processed as described in methods, a name search through the chronogram 269 database can be performed in less than a minute, even with a very large number of taxon 270 names (Fig. 1). 271

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 288 sample includes two species of cardinals: the black-thighed grosbeak – Pheucticus tibialis 289 and the crimson-collared grosbeak – Rhodothraupis celaeno; three species of buntings: the 290 yellowhammer – Emberiza citrinella, the pine bunting – Emberiza leucocephalos and the 291 vellow-throated bunting – Emberiza elegans; and one species of tanager, the vegetarian finch – 292 Platyspiza crassirostris. Processing of input names found that Emberiza elegans is synonym 293 for Schoeniclus elegans in the default reference taxonomy (OTT v3.3, June 1, 2021). For a 294 detailed discussion on the state of the synonym, refer to Avibase (Avibase, 2022; Lepage, 295 2004; Lepage, Vaidya, & Guralnick, 2014). Discovering this synonym allowed assigning five 296 age data points for the parent node of Emberiza elegans, shown as Schoeniclus elegans in 297 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 307 topology for these taxa (Fig. 2d), and congruified and mapped age data to nodes in this 308 chosen topology, shown in Table 1. The name processing step allowed including five data 309 points for node "n4" (parent of Schoeniclus elegans) that would not have had any data 310 otherwise due to name mismatch. Age summary statistics per node were calculated (Table 2) 311 and used as calibrations to date the tree topology using the BLADJ algorithm. As expected, 312 more inclusive nodes (e.g., node "n1") have more variance in age data than less inclusive 313 nodes (e.g., node "n5"). Median summary age data for node "n2" was excluded as final 314 calibration because it is older than the median age of a more inclusive node, "n1" (Fig. 2g). 315

## An Example with the Family of True Finches

Creating a query. To obtain ages for all species within the family of true finches, 317 Fringillidae, we ran a DateLife query using the "get species from taxon" flag, which gets all 318 recognized species names within a named group from a taxonomy of choice. Following the 319 NCBI taxonomy, our DateLife query has 289 Fringillidae species names. This 320 taxon-constrained approach implies that the full DateLife analysis will be performed using a 321 tree topology and ages available for species names from a given taxonomic group, which do 322 not necessarily correspond to a monophyletic group. Users can change this behavior by 323 providing all species names corresponding to a monophyletic group as input for a DateLife 324 search, or a monophyletic tree to construct a DateLife summary. 325

**Searching the database.** Next, we used the processed species names in our 326 DateLife query to identify chronograms with at least two Fringillidae species as tip taxa. 327 The DateLife search identified 19 chronograms matching this criteria, published in 13 328 different studies (Barker et al., 2013, 2015; Burns et al., 2014; Claramunt & Cracraft, 2015; 329 Gibb et al., 2015; Hedges et al., 2015; Hooper & Price, 2017; Jetz et al., 2012; Kimball et al., 330 2019; Oliveros et al., 2019; Price et al., 2014; Roquet, Lavergne, & Thuiller, 2014; Uyeda, 331 Pennell, Miller, Maia, & McClain, 2017). Once identified, DateLife pruned these matching 332 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 345 OpenTree's topology shown in Supplementary Figure S2, reducing the original 1,206 node 346 age data set to 818 different data points (Supplementary Table S2) that could be used as 347 calibrations for that chosen topology. The congruent node age data points were summarized 348 for each node, resulting in 194 summary node ages. From these 21 were excluded as 349 secondary calibrations because they were older than the ancestral node. The remaining 173 350 summary node ages were used as secondary calibrations to obtain a fully dated (and 351 resolved) phylogeny with the program BLADJ (Supplementary Figure S3). Results for a 352 subgroup are shown in Figure 3. 353

#### CROSS-VALIDATION TEST

We performed a cross validation test of a DateLife analysis using the Fringillidae 355 source chronograms obtained above (Supplementary Figs. S5-S23). As inputs for a DateLife 356 analysis, we used all individual tree topologies from each of the 19 source chronograms from 357 13 studies, treating their node ages as unknown. We congruified node ages extracted from 358 chronograms from all other studies upon the individual topologies, effectively excluding 359 original ages from each topology. Finally, average node ages per node were applied as 360 secondary calibrations and smoothed with the BLADJ algorithm. We found that node ages 361 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 365 are generally older in the original study than estimated using cross-validated ages 366 (Supplementary Fig. S4). In general, topologically deeper nodes display the largest age 367

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

DateLife.

DISCUSSION

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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. 377 Incorporation of more chronograms into the database will continue to improve DateLife's 378 services. One option to increase the number of chronograms in the DateLife database is the 379 Dryad data repository. Methods to automatically mine chronograms from Dryad could be 380 designed and implemented. However, Dryad's metadata system has no information to 381 automatically detect branch length units, and those would still need to be determined 382 manually by a human curator. We would like to emphasize on the importance of sharing 383 chronogram data, including systematically curated metadata, into open repositories, such as OpenTree's Phylesystem (McTavish et al., 2015) for the benefit of research and the scientific 385 community as a whole. Another important source of expert data on time of lineage 386 divergence is TimeTree's database (Hedges, Dudley, & Kumar, 2006), which holds 387 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 390 website application were updated, now allowing some kinds of reuse, but not redistribution. 391 The inaccessibility of TimeTree's database was an inspiration for the DateLife project, 392

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 397 beyond, we emphasize that DateLife's results –as well as any insights gleaned from them, 398 largely depend on the quality of the source chronograms: low quality chronograms will 390 produce low quality results. The "garbage in, garbage out" problem has long been recognised 400 in supertree methods for summarizing phylogenetic trees (Bininda-Emonds et al., 2004). We 401 note that this is a surfacing issue of any automated tool for biological data analysis. For 402 example. DNA riddled with sequencing errors will produce generally poor alignments that 403 will return biased evolutionary hypothesis, independently of the quality of the analysis 404 software used. Again, we urge readers and DateLife users to explore all input chronograms 405 before using a summary chronogram resulting from a DateLife workflow.

Finally, uncertainty and variability of chronogram node age estimates might pose larger issues in some research areas than others. For example, in ecological and conservation biology studies, it has been shown that incorporating some chronogram data provides better results than when not using any age data at all, even if the node ages are not 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 419 available for the same taxa have potentially been estimated implementing different types of 420 calibrations, which affects the resulting node age estimates. For example, in the DateLife 421 analysis of the Fringillidae shown above, the chronograms from one study (Burns et al., 2014) 422 were inferred using molecular substitution rate estimates across birds (Weir & Schluter, 423 2008), and have much older age estimates for the same nodes than chronograms that were 424 inferred using fossil calibrations (Figs. 3, 4c, Supplementary Figs. S4c, S10). Another source 425 of conflict in estimated node ages can arise from different placements for the same calibration, 426 which would imply fundamentally distinct evolutionary hypotheses (Antonelli et al., 2017). 427 For example, two independent researchers working on the same clade should both carefully 428 select and justify their choices of fossil calibration placement. Yet, if one researcher concludes 429 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 resulting age estimates will 431 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 433 also affects uncertainty of node age estimates. For example, nodes that are sandwiched 434 between a calibrated node and a calibrated root have less freedom of movement and hence 435 narrower confidence intervals (Vos & Mooers, 2004), which inflates precision for nodes 436 without calibrations but does not necessarily improve accuracy of the estimated ages. 437

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 445 chronograms, i.e. secondary calibrations. In general, the scientific community has more 446 confidence in chronograms using primary calibrations, where the dated tree is generated from 447 a single analysis where carefully chosen fossil calibrations are the source of absolute time 448 information, than in analyses dated using secondary calibrations (Antonelli et al., 2017; 449 Garzón-Orduña, Silva-Brandão, Willmott, Freitas, & Brower, 2015; Graur & Martin, 2004; 450 Sauguet, 2013; Sauguet et al., 2012; Schenk, 2016; Shaul & Graur, 2002). However, 451 implementation of primary calibrations is difficult: it requires specialized expertise and 452 training to discover, place and apply calibrations appropriately (Hipsley & Müller, 2014; 453 Ksepka et al., 2011). One approach is to use fossils that have been widely discussed and 454 previously curated as calibrations to date other trees (Ksepka et al., 2011; Sauquet, 2013), 455 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., 459 2015). This database facilitates the use of expert primary fossil calibrations in new 460 phylogenetic dating analyses. Yet, users still require the expertise to locate and calibrate 461 appropriate nodes in their phylogenies which correspond with fossils available in the 462 database. 463

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 slightly older when estimated using our secondary calibrations than ages from the original 473 studies. The only exception to this trend was observed in Burns et al. (2014) chronogram, 474 which generally displays much younger node ages when estimated using secondary 475 calibrations than the original study (Supplementary Figs. S4c, S10), supporting previous 476 observations (Sauquet et al., 2012; Schenk, 2016). However, these younger dates are more 477 likely an example of how multiple secondary calibrations can correct erroneous estimates, as 478 dates on the Burns et al. (2014) tree were obtained using a single secondary calibration 479 based on a previously estimated molecular evolution rate across birds from Weir and 480 Schluter (2008), and appear as major outliers compared to alternate estimates for the same 481 nodes based on primary fossil calibrations (Fig. 3, Supplementary Fig. S3). 482

# 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.

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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 503 resource in research, as it provides the opportunity to work with a chronogram that reflects a 504 unified evolutionary history for a lineage, by putting together evidence from different 505 hypotheses. For example, the largest, and taxonomically broadest chronogram currently 506 available from OpenTree was constructed summarizing age data from 2,274 published 507 chronograms using NCBI's taxonomic tree as backbone (Hedges et al., 2015), which has been 508 widely reused for research. Finally, we note that summarizing chronograms should be done 509 with caution, as it may amplify the effect of uncertainty and errors in source data, and blur 510 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). 512

# 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 524 was simulated; Rabosky et al. (2018) created a chronogram of 31,536 ray-finned fishes, of 525 which only 37% had molecular data; Smith and Brown (2018) constructed a chronogram of 526 353,185 seed plants where only 23% had molecular data. Stochastically resolved chronograms 527 can return diversification rates estimates that appear less biased than those estimated from 528 their incompletely sampled counterparts, even with methods that account for missing 520 lineages by using sampling fractions (Chang et al., 2020; Cusimano et al., 2012), but can also 530 introduce spurious patterns of early bursts of diversification (Cusimano & Renner, 2010; Sun 531 et al., 2020). 532

Taxonomy-based stochastic polytomy resolvers also introduce topological differences in 533 phylogenetic trees. The study of macroevolutionary processes largely depends on an 534 understanding of the timing of species diversification events, and different phylogenetic and 535 chronogram hypothesis can provide very different overviews of the macroevolutionary history 536 of a biological group. For example, alternative topologies in chronograms from the same 537 biological group can infer very different species diversification patterns (Rabosky, 2015; Title 538 & Rabosky, 2016). Similarly, there are worries that patterns of morphological evolution 539 cannot be accurately inferred with phylogenies that have been resolved stochastically over a 540 taxonomic backbone, as any patterns would be erased by randomization (Rabosky, 2015). 541 We note that the same applies for geography- and morphology-dependent diversification 542 analysis. Hence, we suggest that phylogenies that have been processed with taxonomy-based 543 stochastic polytomy resolvers, including certain summary chronograms from a DateLife 544 analysis, can be useful as null or neutral models, representing the case of a diversification 545 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 555

Knowledge of the evolutionary time frame of organisms is key to many research areas: 556 trait evolution, species diversification, biogeography, macroecology and more. It is also 557 crucial for education, science communication and policy, but generating chronograms is 558 difficult, especially for those who want to use phylogenies but who are not systematists, or 559 do not have the time to acquire and develop the necessary knowledge and skills to construct 560 them on their own. Importantly, years of primarily publicly funded research have resulted in 561 vast amounts of chronograms that are already available in scientific publications, but 562 functionally hidden from the public and scientific community for reuse. 563

The DateLife project allows for easy and fast summarization of public and 564 state-of-the-art data on time of lineage divergence. It is available as an R package, and as a 565 web-based R shiny application at www.datelife.org. DateLife provides a straightforward way 566 to get an informed picture of the state of knowledge for the time frame of evolution of 567 different regions of the tree of life, and allows identifying regions that require more research, or that have conflicting information. Additionally, both summary and newly generated trees using the DateLife workflow are useful to evaluate evolutionary hypotheses in different areas of research. We hope that the DateLife project will increase awareness of the existing variation in expert estimations of time of divergence, and foster exploration of the effect of 572 alternative divergence time hypotheses on the results of analyses, nurturing a culture of more 573 cautious interpretation of evolutionary results.

## 575 AVAILABILITY

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The DateLife software is free and open source. It can be used online through its R 576 shiny web application hosted at http://www.datelife.org, and locally through the datelife 577 R package, available from Zenodo (https://doi.org/10.5281/zenodo.593938 and the CRAN 578 repository (Sanchez-Reves et al., 2022). DateLife's web application is maintained using 570 RStudio's shiny server and the shiny package open infrastructure, as well as Docker and 580 OpenTree's infrastructure (datelife.opentreeoflife.org). datelife's stable version can be 581 installed from the CRAN repository using the command install.packages(pkgs = 582 "datelife") from within R. Development versions are available from DateLife's GitHub 583 repository (https://github.com/phylotastic/datelife) and can be installed using the 584 command devtools::install github("phylotastic/datelife"). 585

#### SUPPLEMENTARY MATERIAL

Supplementary material, including figures, tables, code, biological examples, 587 benchmark results, data files and online-only appendices, can be viewed and downloaded 588 from the Dryad data repository (https://doi.org/10.5061/dryad.cnp5hqc6w), as well as from 580 the Zenodo stable repositories that host the reproducible manuscript 590 (https://doi.org/10.5281/zenodo.7435094), the biological examples 591 (https://doi.org/10.5281/zenodo.7435101), the software benchmark (https://doi.org/10.5281/zenodo.7435106), and the figures 593 (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, 596 https://github.com/LunaSare/datelife examples, and 597 https://github.com/LunaSare/datelife benchmark. 598

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# **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 950 Passeriformes (a, b). Processed species names were found across 9 chronograms within 6 951 independent studies (c; Barker et al. (2012), Barker et al. (2015), Burns et al. (2014), 952 Hedges et al. (2015), Hooper and Price (2017), Jetz et al. (2012). This revealed 28 source 953 age data points for the queried species names (e; Table 1). Summarized age data (f; Table 2) 954 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). The Paleogene Period spans from the end of the Cretaceous Period 66 million years ago (Ma) 957 to the beginning of the Neogene Period 23.03 Ma. 958

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 253 963 nodes; DateLife revealed age data for 194 of these nodes from at least one published 964 chronogram. In total, 19 different chronograms from 13 different studies contributed 818 age 965 data points, which were summarized to obtain a single value for each one of the 194 nodes 966 with age data. From the 194 summary ages available, 21 were discarded and not used as 967 calibrations (asterisk, \*), because they were older than a parent node or younger than a 968 descendant node; the remaining 173 summary ages were used as secondary calibrations 969 (forward slash, /) with the Branch Length Adjuster (BLADJ) software from Webb et al., 970 (2008). The Paleogene Period spans from the end of the Cretaceous Period 66 million years 971 ago (Ma) to the beginning of the Neogene Period 23.03 Ma. 972

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).