## APPLICATION



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# The PHRUTA R package: Increasing access, reproducibility and transparency in phylogenetic analyses

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### **Abstract**

- 1. Current practices for assembling phylogenetic trees often recur to sequence data stored in GenBank. However, the molecular and taxonomic make-up of sequences deposited in GenBank is generally not very clear.
- 2. PHRUTA, a newly developed R package, is designed to primarily improve access to genetic data stored in GenBank. Functions in PHRUTA further enable users to assemble single- and multi-gene molecular datasets, and run basic phylogenetic tasks, all within R.
- 3. The structure of the functions implemented in PHRUTA aim to allow users to assemble simple workflows for particular tasks, which are in turn expected to increase reproducibility of relatively simple phylogenies.
- 4. To support the use of PHRUTA by researchers in different fields with variable levels of coding expertise, this paper presents the first version of SALPHYCON, a shiny web app that aims to increase access to the fundamental functions in the PHRUTA R package.

### KEYWORDS

Phylogenetics, R package, reproducibility, workflow

### 1 | INTRODUCTION

Although the assemblage of molecular phylogenies has been the backbone of numerous studies, the existing tools used to retrieve sequences from public databases, curate molecular datasets, assemble multi-locus alignments and finally infer phylogenies generally involve an extensive set of software that, in many cases, is poorly streamlined (Hall, 2004; Lemey et al., 2009; Wiley & Lieberman, 2011). In fact, it has been estimated that nearly 60%–70% of phylogenetic analyses are not reproducible (Drew, 2013; Drew et al., 2013; Magee et al., 2014; Stoltzfus et al., 2012) and, consequently, are "lost to science" (Magee et al., 2014). Here, I present PHRUTA, an R package, and SALPHYCON, the associated interactive application, designed to

improve the reproducibility and access to a fraction of the existing tools of the phylogenetic workflow.

Although similar functions for assembling curated molecular datasets for phylogenetic analyses can be found in PHYLOTAR (Wang et al., 2022) and SUPERCRUNCH (Portik & Wiens, 2020), PHRUTA simplifies these functions into a single open-source package that uses the numerous required tools. For example, PHYLOTAR is limited to downloading and curating sequences (e.g. it does not align sequences) and, SUPERCRUNCH curates sequences that are already stored locally. SUPERSMART (Antonelli et al., 2017) and the associated R workflow SUPERSMARTR (Antonelli et al., 2017; Bennett et al., 2018) also contain similar functions, some of which have been simplified here in a single package, PHRUTA. PHRUTA improves upon

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standalone applications MEGA (Tamura et al., 2007) and geneious (Kearse et al., 2012) by decreasing the burden for researchers to perform and publish reproducible computational workflows (e.g. MEGA) and eliminating unnecessary financial barriers for all researchers in performing said analyses (e.g. geneious has a paid version). Finally, note that PHRUTA is already integrated within R (R Core Team, 2022), a language that contains the state of art packages and methodological implementations in phylogenetics and many different fields in biology and related fields (e.g. Revell & Harmon, 2022).

In short, PHRUTA provides an early perspective on how to improve reproducibility and access to the phylogenetic pipeline in R. In this manuscript, I provide a brief overview of PHRUTA, briefly describing its current capabilities. Finally, I present the first version of SALPHYCON, a SHINY app (Chang et al., 2021) designed to even further increase access to reproducible phylogenetic pipelines via PHRUTA, and discuss its functionality and future outlook.

### 2 | THE 'PHRUTA' 'R' PACKAGE

The package PHRUTA is primarily designed to simplify the basic phylogenetic computational workflow, entirely in R. PHRUTA is expected to allow scientists to gather molecular datasets or phylogenies for particular taxonomic groups, with minimal complexity and maximal reproducibility.

In general, Phruta (1) scans GenBank (Benson et al., 2012) primarily using the Rutils (Schöfl, 2016) and Rentrez (Winter, 2017) R packages to identify potential phylogenetically relevant gene regions for a given set of taxa, (2) retrieves gene sequences and curates taxonomic information from different taxonomic backbones, (3) combines downloaded and local gene sequences, and (4) performs phylogenetic tasks such as sequence alignment, phylogenetic inference and basic tree dating. An interactive web application, SALPHYCON, is also available as an additional resource to run the basic functions of PHRUTA without the need for any previous R programming experience.

The current version of PHRUTA includes a set of eight major functions. Depending on the information available, running all eight major functions in PHRUTA potentially results in a time-calibrated phylogeny for a given set of taxa. Note that all the functions for which the primary output is sequences (aligned or unaligned) are listed under sq.\*. All the functions that output phylogenies (time-calibrated or not) are listed under tree.\*.

• First, the distribution of genes sampled in GenBank for a given organism or set of taxa can be explored using the *acc.gene.sam-pling()* function. This function in PHRUTA extends the functionality of the RUTILS R package by scraping data from GenBank. *acc.gene. sampling()* returns a table that sums up the distribution of genes sampled for a given taxon or set of taxa. Note that *acc.gene.sam-pling()* only retrieves a set of gene names but does not download any sequences.

- Second, for a given list of accession numbers for given a list of organisms can be retrieved using the acc.table.retrieve() function.
   Instead of directly downloading sequences from GenBank (see sq.retrieve.direct() below), retrieving accession numbers as an intermediate step allows users to have more control over the sequences that are being used in downstream analyses.
- Third, two different functions in PHRUTA enable users to download GenBank sequences. First, sequences can be downloaded using sq.retrieve.indirect() on the accession numbers retrieved before using the acc.table.retrieve() function presented above. Second, users can also skip defining an accession number table and download gene sequences directly using the sq.retrieve.direct() function in PHRUTA. Note that sq.retrieve.direct() is primarily based on functions from RENTREZ (Winter, 2017) and sq.retrieve.indirect() mainly uses functions from RUTILS (Schöfl, 2016). Optionally, these functions will create a folder 0.Sequences that includes all the retrieved sequences. Note that using acc.table.retrieve() is the preferred option within PHRUTA.
- Fourth, local sequences to those retrieved from GenBank can be included into the workflow using the *sq. add()*. This function saves all resulting *fasta* files in two directories: (1) combined sequences are stored in *0.Sequences* and local sequences are moved into *0.AdditionalSequences*. Note that the originally downloaded sequences are moved to *0.0.OriginalDownloaded* at this step. *sq.add()* is primarily based on functions in the APE R package (Paradis & Schliep, 2019).
- Fifth, the *sq.curate()* function filters out unreliable sequences based on information listed in GenBank (e.g. the prefix *PREDICTED*) and on the target taxonomic information provided by the user. For instance, if a given species belongs to a non-target group, this species is dropped from the analyses. *sq.curate()* function automatically corrects taxonomy based on alternative taxonomic databases, renames sequences, and further uses the ODSEQ R package (Jiménez, 2022) to detect outliers in the set of sequences.
- Sixth, sq.aln() performs multiple sequence alignment in fasta files.
   PHRUTA uses the DECIPHER R package for this purpose (Wright, 2016, 2020), which allows for adjusting sequence orientation and masking (removing ambiguous sites).
- Seventh, the tree.raxml() function allows users to perform tree inference under RAxML for sequences in a given folder. This is a wrapper to ips::raxml() (Heibl, 2008) and each of the arguments can be customized. The current release of PHRUTA can manage both partitioned and unpartitioned analyses in RAxML (Stamatakis, 2014). Note that users can also provide starting and constrained trees.
- Eight, tree.dating() enables users to perform time-calibrations of a given phylogeny (Eastman et al., 2013) using geiger::congruify.phylo() (Harmon et al., 2008; Pennell et al., 2014). PHRUTA includes a basic set of comprehensively sampled, time-calibrated phylogenies that are used to extract secondary calibrations for the target phylogeny. Users can choose to run either PATHD-8 (Britton et al., 2007) or TREEPL (Smith & O'Meara, 2012) for the calibration step.

# 3 | BRIEF TUTORIAL: USING 'PHRUTA' TO INFER THE PHYLOGENETICS OF NEW WORD QUAILS

Let us learn how PHRUTA works by assembling a molecular dataset at the species level for a handful of bird genera. Note that although this tutorial is based on a particular set of taxa, users can decide on what can choose their target clades in other families, orders, or even kingdoms.

Here, we will focus on assembling a phylogeny for the new-world quail (Johnsgard, 1988). Species in this group are classified in the family Odontophoridae, a clade that includes nearly 34 extant species classified in 10 genera. In general, the higher-level taxonomic information in GenBank for the Odontophoridae is largely congruent with recent studies on the systematics of the group (Cohen et al., 2012; Crowe, Bloomer, et al., 2006; Crowe, Bowie, et al., 2006; Hosner et al., 2015). However, GenBank classifies Ptilopachus, a genus commonly included under the Odontophoridae, as part of the Phasianidae. We will follow more recent studies suggesting that Ptilopachus is nested within the Odontophoridae. As our outgroup, we will select the Phasianidae. Within this later family, Phasianidae, we will explicitly focus on sampling species in the genus Polyplectron, a clade of eight extant species. Finally, given that the systematics of the Odontophoridae has been discussed before using morphological and molecular evidence, we will be able to compare the topology of our tree with recent studies (Cohen et al., 2012; Crowe, Bloomer, et al., 2006; Crowe, Bowie, et al., 2006; Hosner et al., 2015).

Up to this point, we have decided on the taxonomic makeup of our analyses. From here, we could simply check the molecular sampling used in previous studies and search for those genes in GenBank for Odontophoridae and *Polyplectron* (Cohen et al., 2012; Crowe, Bloomer, et al., 2006; Crowe, Bowie, et al., 2006; Hosner et al., 2015). Alternatively, we could use PHRUTA to determine genes well sampled in GenBank for both the ingroup and outgroup. For simplicity, we will follow the latter procedure by using the *gene*. *sampling.retrieve()* function in PHRUTA. An object of class data.frame named gs.seqs is generated containing the names of different gene regions that are sampled in GenBank for the target taxa.

Given the search terms, PHRUTA retrieved the names for ~80 gene regions from GenBank. Note that the *gene.sampling.retrieve()* function provides an estimate of the number of species in GenBank that match the taxonomic criteria of the search term and that have sequences for a given gene region. However, the distribution of gene names and their species-level coverage is only as good as the annotations for genes deposited in GenBank. Estimates of species sampled within genes could be strongly affected in some clades where

TABLE 1 Top six gene regions sampled in GenBank for the Odontophoridae and *Polyplectron*. In the table, the number of species sampled for a given gene region is shown along with the relative number of species in GenBank across the examined genes. The number of species per gene region shown in this table is as good as the taxonomic information deposited in GenBank. Note that PHRUTA was able to retrieve the names of ~80 gene regions but only a summary of them are shown in the table.

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Gene region	Sampled in # of species (NCBI taxonomy)	Percent of sampled species
NADH dehydrogenase subunit 2	92	99
12S ribosomal RNA	28	30
eukaryotic elongation factor 2	26	28
NADH dehydrogenase subunit 5	19	20
cytochrome b	16	17
cytochrome oxidase subunit 1	10	11

information on GenBank is highly incomplete or unreliable. I show a summary of the resulting search in Table 1.

Using the combination of well-sampled genes and our list of taxa, we will now generate a preliminary summary of the accession numbers. I call this dataset preliminary because not all these accession numbers are expected to be in the final molecular dataset. For instance, some sequences may be removed after taxonomic synonyms are identified in the dataset.

From this point, we will assemble a species-level summary of accession numbers using the *acc.table.retrieve()* function in PHRUTA (i.e. *speciesLevel=TRUE'* argument). For simplicity, this tutorial will focus on analysing gene regions that are sampled in >20% of the species (*targetGenes* data.frame). We use the line below to select genes based on a given threshold of interest (e.g. 20 reflecting 20%; this can be modified by the user).

```
```{r}
targetGenes <- gs.seqs[gs.seqs$PercentOfSampledSpecies > 20,]
...
```

The *acc.table* object created below is a data.frame object that will later be used to download the relevant gene sequences from GenBank (see Table 2).

TABLE 2 Accession numbers for the Odontophoridae and *Polyplectron* obtained using PHRUTA. We show the species-level sampling based on gene regions with >20% of species (see Table 1).

Accession number	Species	Gene region
MZ476322	Callipepla californica	NADH dehydrogenase subunit 2
MZ476314	Callipepla gambelii	NADH dehydrogenase subunit 2
EU166949	Colinus virginianus	NADH dehydrogenase subunit 2
AF222544	Colinus cristatus	NADH dehydrogenase subunit 2
KR732857	Colinus nigrogularis	NADH dehydrogenase subunit 2
KR732856	Dendrortyx barbatus	NADH dehydrogenase subunit 2
KR732855	Philortyx fasciatus	NADH dehydrogenase subunit 2
KR732850	Rhynchortyx cinctus	NADH dehydrogenase subunit 2
KR732849	Cyrtonyx montezumae	NADH dehydrogenase subunit 2
KR732848	Oreortyx pictus	NADH dehydrogenase subunit 2
KR732847	Odontophorus leucolaemus	NADH dehydrogenase subunit 2
KR732846	Odontophorus speciosus	NADH dehydrogenase subunit 2
KR732845	Odontophorus erythrops	NADH dehydrogenase subunit 2
KR732844	Odontophorus guttatus	NADH dehydrogenase subunit 2
KR732843	Odontophorus gujanensis	NADH dehydrogenase subunit 2
KR732842	Odontophorus capueira	NADH dehydrogenase subunit 2
KR732841	Odontophorus stellatus	NADH dehydrogenase subunit 2
KR732840	Dactylortyx thoracicus	NADH dehydrogenase subunit 2
KR732839	Dendrortyx macroura	NADH dehydrogenase subunit 2
KR732838	Callipepla squamata	NADH dehydrogenase subunit 2
KR732837	Callipepla douglasii	NADH dehydrogenase subunit 2
KC556543	Colinus leucopogon	NADH dehydrogenase subunit 2
KC556066	Dendrortyx leucophrys	NADH dehydrogenase subunit 2
KC556060	Cyrtonyx ocellatus	NADH dehydrogenase subunit 2
KC556524	Odontophorus balliviani	NADH dehydrogenase subunit 2

### TABLE 2 (Continued)

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Accession number	Species	Gene region
KC556517	Odontophorus columbianus	NADH dehydrogenase subunit 2
KC556515	Odontophorus strophium	NADH dehydrogenase subunit 2
KC556513	Odontophorus melanonotus	NADH dehydrogenase subunit 2
KC556512	Odontophorus hyperythrus	NADH dehydrogenase subunit 2
KC556507	Odontophorus melanotis	NADH dehydrogenase subunit 2
DQ768288	Francolinus nahani	NADH dehydrogenase subunit 2
KR732851	Ptilopachus petrosus	NADH dehydrogenase subunit 2
EF569482	Polyplectron inopinatum	NADH dehydrogenase subunit 2
EF569481	Polyplectron napoleonis	NADH dehydrogenase subunit 2
EF569480	Polyplectron chalcurum	NADH dehydrogenase subunit 2
EF569479	Polyplectron bicalcaratum	NADH dehydrogenase subunit 2
DQ768268	Polyplectron malacense	NADH dehydrogenase subunit 2
DQ768266	Polyplectron germaini	NADH dehydrogenase subunit 2
KC778823	Polyplectron katsumatae	NADH dehydrogenase subunit 2
KR732830	Cyrtonyx montezumae	12S ribosomal RNA
KR732829	Dactylortyx thoracicus	12S ribosomal RNA
KR732828	Oreortyx pictus	12S ribosomal RNA
KR732827	Odontophorus erythrops	12S ribosomal RNA
KR732826	Odontophorus gujanensis	12S ribosomal RNA
KR732825	Odontophorus stellatus	12S ribosomal RNA
KR732824	Odontophorus capueira	12S ribosomal RNA
KR732823	Odontophorus speciosus	12S ribosomal RNA
KR732822	Odontophorus leucolaemus	12S ribosomal RNA
KR732821	Odontophorus balliviani	12S ribosomal RNA
KR732820	Dendrortyx macroura	12S ribosomal RNA
KR732819	Philortyx fasciatus	12S ribosomal RNA
KR732818	Callipepla gambelii	12S ribosomal RNA
KR732817	Callipepla californica	12S ribosomal RNA
KR732816	Callipepla douglasii	12S ribosomal RNA
KR732815	Callipepla squamata	12S ribosomal RNA
KR732814	Colinus cristatus	12S ribosomal RNA
KR732813	Colinus nigrogularis	12S ribosomal RNA

TABLE 2 (Continued)

Accession number	Species	Gene region
KR732812	Colinus virginianus	12S ribosomal RNA
KR732832	Ptilopachus petrosus	12S ribosomal RNA
KC749467	Polyplectron inopinatum	12S ribosomal RNA
KC749466	Polyplectron germaini	12S ribosomal RNA
KC749465	Polyplectron napoleonis	12S ribosomal RNA
KC749464	Polyplectron chalcurum	12S ribosomal RNA
KC749463	Polyplectron bicalcaratum	12S ribosomal RNA
KC778974	Polyplectron katsumatae	12S ribosomal RNA
KR732895	Colinus cristatus	eukaryotic elongation factor 2
KR732894	Callipepla squamata	eukaryotic elongation factor 2
KR732893	Dendrortyx macroura	eukaryotic elongation factor 2
KR732892	Philortyx fasciatus	eukaryotic elongation factor 2
KR732891	Cyrtonyx montezumae	eukaryotic elongation factor 2
KR732889	Rhynchortyx cinctus	eukaryotic elongation factor 2
KR732888	Oreortyx pictus	eukaryotic elongation factor 2
KR732887	Colinus virginianus	eukaryotic elongation factor 2
KR732886	Callipepla douglasii	eukaryotic elongation factor 2
KR732885	Callipepla californica	eukaryotic elongation factor 2
KR732884	Callipepla gambelii	eukaryotic elongation factor 2
KR732883	Odontophorus erythrops	eukaryotic elongation factor 2
KR732882	Odontophorus guttatus	eukaryotic elongation factor 2
KR732881	Odontophorus speciosus	eukaryotic elongation factor 2
KR732880	Odontophorus leucolaemus	eukaryotic elongation factor 2
KR732879	Odontophorus balliviani	eukaryotic elongation factor 2
KR732878	Odontophorus stellatus	eukaryotic elongation factor 2
KR732877	Odontophorus capueira	eukaryotic elongation factor 2
KR732890	Ptilopachus petrosus	eukaryotic elongation factor 2
KC749707	Polyplectron inopinatum	eukaryotic elongation factor 2

TABLE 2 (Continued)

ABLE 2 (Co	illilided)	
Accession number	Species	Gene region
KC749706	Polyplectron germaini	eukaryotic elongation factor 2
KC749705	Polyplectron chalcurum	eukaryotic elongation factor 2
KC749704	Polyplectron bicalcaratum	eukaryotic elongation factor 2
KC778867	Polyplectron katsumatae	eukaryotic elongation factor 2
KR732875	Philortyx fasciatus	NADH dehydrogenase subunit 5
KR732874	Callipepla douglasii	NADH dehydrogenase subunit 5
KR732872	Rhynchortyx cinctus	NADH dehydrogenase subunit 5
KR732871	Dactylortyx thoracicus	NADH dehydrogenase subunit 5
KR732870	Odontophorus leucolaemus	NADH dehydrogenase subunit 5
KR732869	Odontophorus balliviani	NADH dehydrogenase subunit 5
KR732868	Odontophorus guttatus	NADH dehydrogenase subunit 5
KR732867	Odontophorus erythrops	NADH dehydrogenase subunit 5
KR732866	Odontophorus capueira	NADH dehydrogenase subunit 5
KR732865	Odontophorus stellatus	NADH dehydrogenase subunit 5
KR732864	Odontophorus gujanensis	NADH dehydrogenase subunit 5
KR732863	Dendrortyx macroura	NADH dehydrogenase subunit 5
KR732862	Callipepla californica	NADH dehydrogenase subunit 5
KR732861	Callipepla gambelii	NADH dehydrogenase subunit 5
KR732860	Callipepla squamata	NADH dehydrogenase subunit 5
KR732859	Colinus virginianus	NADH dehydrogenase subunit 5
KR732858	Colinus nigrogularis	NADH dehydrogenase subunit 5
KR732873	Ptilopachus petrosus	NADH dehydrogenase subunit 5

Since we are interested in retrieving sequences from GenBank using the accession numbers dataset generated above, we will use the *sq.retrieve.indirect()* function in PHRUTA. Please note that there are two versions of *sq.retrieve.\** in PHRUTA. The one that we are using in this tutorial, *sq.retrieve.indirect()*, retrieves sequences "indirectly" because it requires a table of accession numbers to be passed as an argument (see the *acc.table.retrieve()* function above). I present

the information in this tutorial using <code>sq.retrieve.indirect()</code> instead of <code>sq.retrieve.direct()</code> because <code>sq.retrieve.indirect()</code> is more flexible and robust. Specifically, <code>sq.retrieve.indirect()</code> allows users to correct issues prior to downloading/retrieving the sequences. For instance, you can add new sequences, species, populations to the resulting data.frame from <code>acc.table.retrieve()</code>. You could even manually assemble your own dataset of accession numbers to be retrieved using <code>sq.retrieve.indirect()</code>. Instead, <code>sq.retrieve.direct()</code> does its best to directly retrieve GenBank sequences for a target set of taxa and set of gene regions. In short, you should be able to catch errors using <code>sq.retrieve.indirect()</code> but mistakes will be harder to spot and fix if you are using <code>sq.retrieve.direct()</code>.

We still need to retrieve all the sequences from the accessions table generated using acc.table.retrieve(). Note that since we have specified download.sqs = FALSE in sq.retrieve.indirect(), the sequences retrieved from GenBank are returned in a list and saved as sqs. downloaded in the R global workspace. If we were to download the sequences to our working directory using the download.sqs = TRUE argument, PHRUTA would have written all the resulting fasta files into a newly created folder named 0.Sequences located in our working directory. If you were interested in exploring more about the export functionalities of PHRUTA, the "to export or not export PHRUTA outputs" vignette in the R package presents a detailed discussion around this aspect.

```
```{r}
sqs.downloaded <- sq.retrieve.indirect(acc.table = acc.table, down-
load.sqs = FALSE)
...</pre>
```

Now, let us make sure that we are only including sequences that are reliable and from species that we are actually interested in analysing. We are going to use the *sq.curate()* function for this. We will provide a list of taxonomic names to filter out incorrect sequences (*filterTaxonomicCriteria* argument). For instance, we could simply provide a vector of the genera that we are interested in analysing. This vector must have a length of 1, with all the target genera being separated with | (e.g. "Callipepla|Colinus|Dendrortyx" if we were interested in only those three genera). For now, we will assume that all of the species we downloaded are relevant to the analyses (i.e. *filterTaxonomicCriteria* = [AZ]). Finally, since we are not downloading anything to our working directory, we need to pass our downloaded sequences (*sqs.downloaded* object generated above using the *sq.retrieve.indirect()* function) to the *sqs.object* argument in *sq.curate()*.

```
'``{r}
sqs.curated <- sq.curate(filterTaxonomicCriteria = '[AZ]', kingdom =
'animals', sqs.object = sqs.downloaded, removeOutliers = FALSE)
...</pre>
```

Running the *sq.curate()* function will create an object of class *list* (i.e. *sqs.curated*) that includes (1) the curated sequences with original names, (2) the curated sequences with species-level names

(renamed\_\* prefix), (3) the accession numbers table (AccessionTable; Table 2) and (4) a summary of taxonomic information for all the species sampled in the files (Table 3). From here, we will align the sequences that we just curated using sq.aln() under default parameters. The sq.aln() function uses the alignment routines implemented in DECIPHER (Wright, 2016, 2020) We are again passing the output from sq.curate(), the object sqs.curated, using the sqs.object argument in sq.aln().

```
```{r}
sqs.aln <- sq.aln(sqs.object = sqs.curated)
...</pre>
```

The resulting multiple sequence alignments will be saved to the *sqs.aln* object in our workspace, a list of alignments. For each of the gene regions, we will have access to the original alignment (*Aln.Original*), the masked one (*Aln.Masked*), and information on the masking process. The masking process is conducted using the *DECIPHER::RemoveGaps()* function by removing positions where gaps are common to all sequences. The raw and masked alignments are presented in Figures 1 and 2, respectively. Note that although the resulting sequences are for now stored in the R workspace, they can be exported to fasta or nexus files using loops or similar functions in R (shown below). The exported sequences (e.g. in *fasta*) can be then used for subsequent analyses in external software.

So far, we have generated a molecular dataset for the target clade. Let us now review how to conduct basic phylogenetic inference in PHRUTA. Phylogenetic inference in PHRUTA is conducted using the tree.raxml() function, a wrapper for the IPS R package. To use this function, we will have to export our sequence alignments locally. We will follow the same folder structure as if we were exporting everything locally from PHRUTA in each step. Specifically, our sequence alignments must be located in a folder named 2.Alignments within our working directory. For this tutorial, we will only export the alignments that were masked. We would not need to generate this folder manually if we were already exporting the results from previous functions in PHRUTA.

```
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```

We are now ready to run RAXML (v.8.2 for this tutorial). We provide details on how to instal RAXML directly on our GitHub repository (https://github.com/ropensci/phruta/blob/main/README.md).

TABLE 3 Summary of taxonomic information for the species sampled for Odontophoridae and *Polyplectron* using PHRUTA. Taxonomic data follows that from the gbif taxonomic backbone for the target species retrieved based on Table 2.

Species name	Genus	Family	Order	Class	Phylum	Kingdom
Callipepla californica	Callipepla	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Callipepla douglasii	Callipepla	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Callipepla gambelii	Callipepla	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Callipepla squamata	Callipepla	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Colinus cristatus	Colinus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Colinus leucopogon	Colinus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Colinus nigrogularis	Colinus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Colinus virginianus	Colinus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Cyrtonyx montezumae	Cyrtonyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Cyrtonyx ocellatus	Cyrtonyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Dactylortyx thoracicus	Dactylortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Dendrortyx barbatus	Dendrortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Dendrortyx leucophrys	Dendrortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Dendrortyx macroura	Dendrortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Francolinus nahani	Francolinus	Phasianidae	Galliformes	Aves	Chordata	Animalia
Odontophorus balliviani	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus capueira	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus columbianus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus erythrops	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus gujanensis	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus guttatus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus hyperythrus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus leucolaemus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus melanonotus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus melanotis	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus speciosus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus stellatus	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Odontophorus strophium	Odontophorus	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Oreortyx pictus	Oreortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Philortyx fasciatus	Philortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia
Polyplectron bicalcaratum	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Polyplectron chalcurum	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Polyplectron germaini	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Polyplectron inopinatum	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Polyplectron katsumatae	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Polyplectron malacense	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Polyplectron napoleonis	Polyplectron	Phasianidae	Galliformes	Aves	Chordata	Animalia
Ptilopachus petrosus	Ptilopachus	Phasianidae	Galliformes	Aves	Chordata	Animalia
Rhynchortyx cinctus	Rhynchortyx	Odontophoridae	Galliformes	Aves	Chordata	Animalia

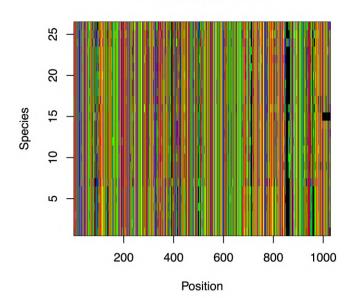
500

### NADH dehydrogenase subunit 2

# Species 10 20 30 1 1 1 1

1000

### 12S ribosomal RNA

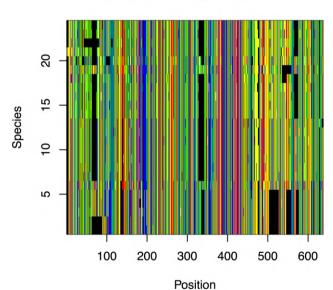


### eukaryotic elongation factor 2

Position

1500

2000



### NADH dehydrogenase subunit 5

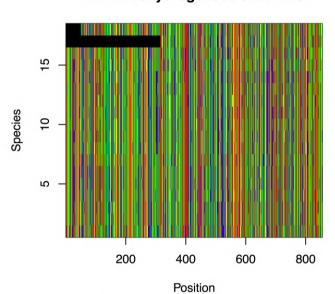


FIGURE 1 Sequence alignments for the Odontophoridae and *Polyplectron*, the selected outgroup, generated using PHRUTA. Highly ambiguous sites have not been masked in this figure. Sequence alignments are shown for four genes that were identified as being represented in >20% of the species sampled in GenBank for the target taxa. In each panel, each column represents a site or position in the alignment. Rows reflect species in the relevant dataset. Grey cells represent gaps in the species/position combination.

2500

Note that in *tree.raxml()*, we will need to indicate where the aligned sequences are located (folder argument), the patterns of the files in the same folder (*FilePatterns* argument; *Masked\_* in our case), and the total of bootstrap replicates. The *outgroup* argument is optional but since we are interested in calibrating our tree afterwards, we will define it using all the species in *Polyplectron*.

```{r}
outgroup <- sqs.curated\$Taxonomy[sqs.curated\$Taxonomy\$genus
== 'Polyplectron',]</pre>

tree.raxml(folder = '2.Alignments',
 FilePatterns = 'Masked\_',
 raxml\_exec = 'raxmlHPC',
 Bootstrap = 100,
 outgroup = paste(outgroup\$species\_names, collapse = ";")
 )

The resulting phylogenetic trees are saved in the 3.Phylogeny folder, also created in our working directory. For many, the

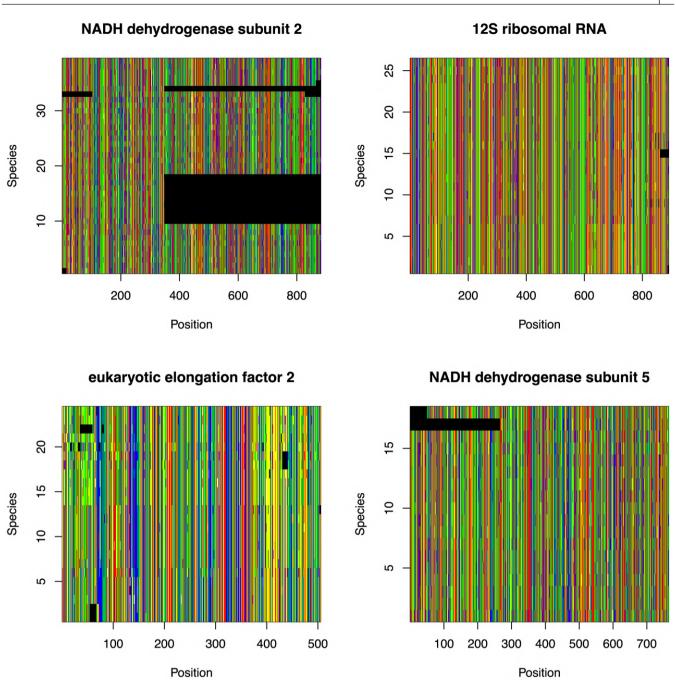


FIGURE 2 Sequence alignments for the Odontophoridae and *Polyplectron*, the selected outgroup, generated using PHRUTA. This figure is equivalent to Figure 1 except that highly ambiguous sites have been masked in this figure. Sequence alignments are shown for four genes that were identified as being represented in >20% of the species sampled in GenBank for the target taxa. In each panel, each column represents a site or position in the alignment. Rows reflect species in the relevant dataset. Grey cells represent gaps in the species/position combination.

bipartitions tree generated in these runs, RAxML\_bipartitions.phruta, are among the most relevant files (Figure 3). The 3.Phylogeny folder further includes the additional RAxML-related input and output files. Note that users can also run partitioned analyses in RAxML within PHRUTA. This approach is available by setting the partitioned argument in tree.raxml() to TRUE. For now, partitioned analyses are based on the gene-based regions that are being analysed. The same model is used to analyse each partition. More details on partitioned analyses can be customized by passing arguments in ips::raxml().

```
tree.raxml(folder = "2.Alignments", FilePatterns = "Masked_",
    raxml_exec = "raxmlHPC", Bootstrap = 100,
    outgroup = paste(outgroup$species_names, collapse = ","),
    partitioned = TRUE
    )
```

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Finally, we note that users will sometimes need to generate constraint trees to seed their searches or limit the tree sampling space.

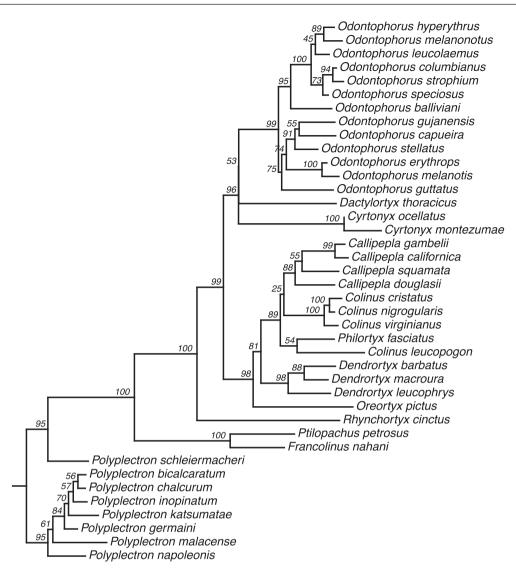


FIGURE 3 Phylogenetic relationships for the Odontophoridae and *Polyplectron*, the selected outgroup, generated using PHRUTA and RAXML. The bipartitions tree is shown in figure, with support values based on a total of 1000 bootstrap replicates. This tree was constructed using a concatenated (but unpartitioned) alignment of all four gene alignments presented in Figure 2.

These constraints can be generated using other software such as TACT (Chang et al., 2019).

### 4 | TREE DATING IN 'PHRUTA'

PHRUTA includes a basic wrapper for functions included in the GEIGER R package that allows for the time-calibration of target phylogenies. Based on the tutorial presented above, we can perform basic tree dating of our phylogeny using secondary calibrations extracted from Scholl and Wiens (2016) phylogeny. I am only using this study because it has a large phylogeny with detailed taxonomic information available for each terminal but I expect to include additional trees in the near future. Users can also choose to use their own reference phylogeny if available. Note that the *tree.dating()*, the function in PHRUTA that is able to conduct phylogenetic dating based on node correspondence between phylogenies, requires the user to specify where

the 1.Taxonomy.csv file is located. This file, 1.Taxonomy.csv, is created automatically when sequences are curated using sq.curate() and results are exported into your local repository. However, since we have been keeping our results in the R global environment, we will have to export 1.Taxonomy.csv manually before we can move forward and time-calibrate the tree.

```{r}
dir.create("1.CuratedSequences")
write.csv(sqs.curated\$Taxonomy, '1.CuratedSequences/1.Taxonomy.
csv')

Tree dating is performed using the *tree.dating()* function in PHRUTA. For this function, we have to provide the name of the folder containing the *1.Taxonomy.csv* file created in *sq.curate()*. We also have to indicate the name of the folder containing the

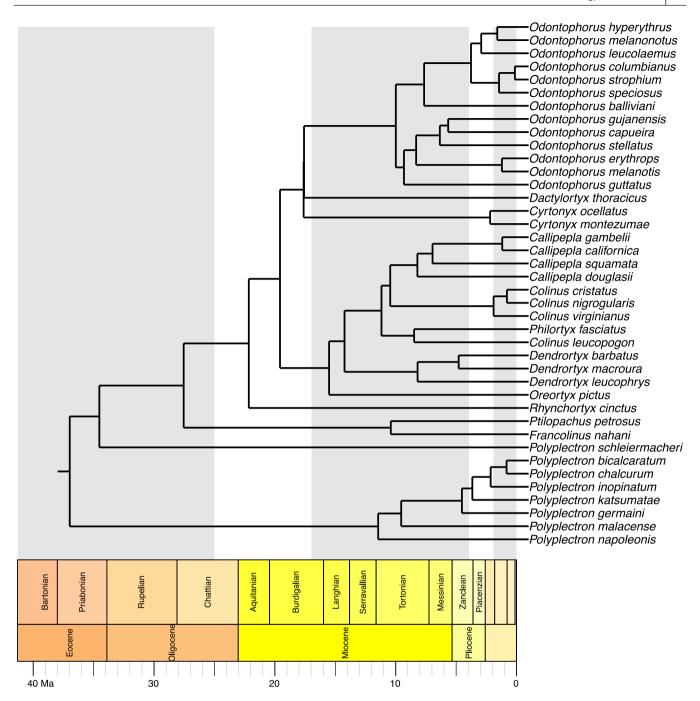


FIGURE 4 Time-calibrated phylogeny for the Odontophoridae and *Polyplectron*, the selected outgroup, generated using PHRUTA and RAXML and secondary calibrations. The backbone topology is presented in Figure 5. The same figure shows support values.

RAxML\_bipartitions.phruta file. We will scale our phylogeny using TREEPL.

resulting time-calibrated tree for the analyses presented in this article is presented in Figure 4.

Running this line will result in a new folder 4.Timetree, which includes the different time-calibrated phylogenies obtained (if any) and associated secondary calibrations used in the analyses. The

### 5 | ADVANCED METHODS WITH 'PHRUTA'

### 5.1 | Curating taxonomic names

You can use *taxonomy.retrieve()*, a function implemented inside *sq.curate()* in PHRUTA to curate species names. For instance, the block of code below will curate taxonomic names using the gbif backbone

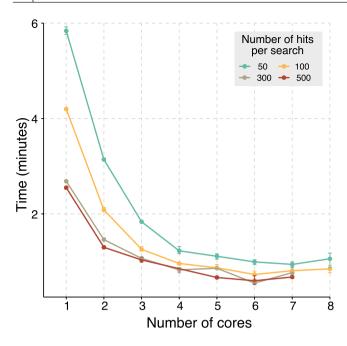


FIGURE 5 Results for performance analyses showing the duration of searches conducted based on the combination of number of hits per search (i.e. number of hits analysed each time the search is conducted) and the number of simultaneous searches conducted from PHRUTA. Note that not all combinations produced results. The search parameters used in this article were optimized based on this analysis.

taxonomy (GBIF Secretariat, 2022). Note that the *kingdom* argument in *taxonomy.retrieve()* can be set to *NULL*, meaning that there will be no indication of the kingdom of the taxa when performing taxonomic searches.

However, GBIF is efficient for retrieving accurate taxonomy when we provide details on the *kingdom*. In the same example, given that all the species we are interested in are animals, we could just use the following block of code to curate taxonomic names.

Depending on your sampling, you could also do the same for plants by using *plants* in the *kingdom* argument instead of *animals*.

Now, what if we were interested in following other databases to retrieve taxonomic information for the species in our database? The latest version of PHRUTA allows users to select the desired database. The databases follow the *taxize::classification()* function (Chamberlain & Szöcs, 2013).

### 5.2 | Running PartitionFinder in 'рнкита'

With the current version of PHRUTA, users are able to run PARTITION FINDER v1 (Lanfear et al., 2012) from R. For this, users should provide the name of the folder where the alignments are stored, a particular pattern in the file names (*Masked\_* in our case), and which models will be run in Partition Finder. This function will download Partition Finder, generate the input files, and run the software, all within R. The output files will be in a new folder within the working directory.

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Unfortunately, the output files are not integrated with the current PHRUTA pipeline. This will be part of a new release. However, users can still perform gene-based partitioned analyses within RAXML or use PartitionFinder's output files to inform their own analyses outside PHRUTA. An upcoming release of PHRUTA will allow users to run more recent versions of PartitionFinder.

### 5.3 | Identifying rogue taxa

PHRUTA can help users run RogueNaRok (Aberer et al., 2011) implemented in the ROGUE R package (Aberer et al., 2013). Users can examine whether rogue taxa should be excluded from the analyses. *tree.roguetaxa()* uses the bootstrap trees generated using the *tree.raxml()* function along with the associated best tree to identify rogue taxa.

```
```{r}
tree.roguetaxa(folder = "3.Phylogeny")
...
```

### 6 | REPRODUCIBILITY WITH 'PHRUTA'

One of the central points for developing PHRUTA was related to increasing the reproducibility of relatively simple phylogenetic analyses. By compiling or calling alternative tools that are

commonly used to assemble species-level molecular and phylogenetic datasets inside an R package, PHRUTA allows users to generate a clear, structured, and reproducible workflow. In fact, PHRUTA is conceived as a package that allows users to choose between at least two alternative options to share their workflow. First, users

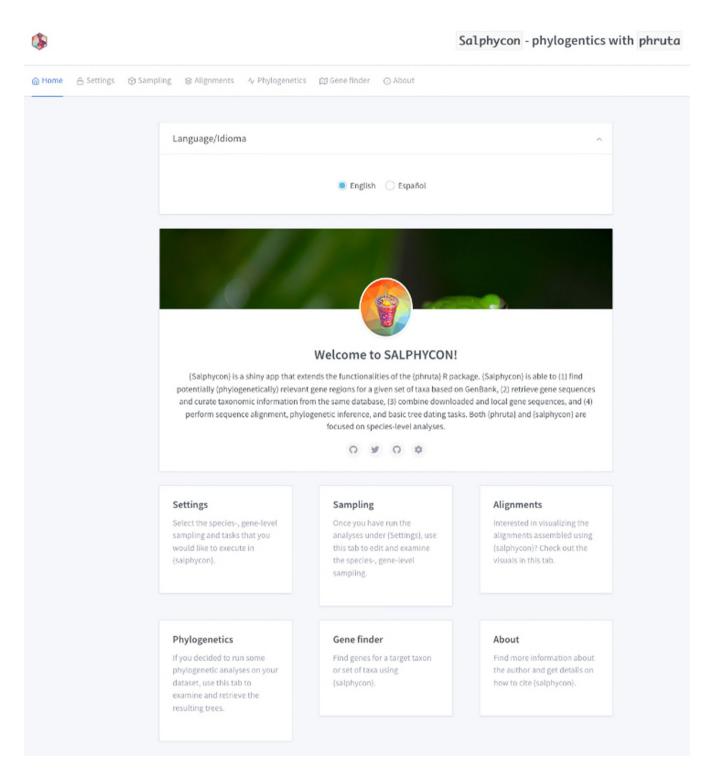


FIGURE 6 Landing page for salphycon, the web application including a set of the basic functions implemented in PHRUTA. In this app, users are able to retrieve sequences from GenBank, explore sequence alignments and conduct basic phylogenetic inference with RAxML. Users can also choose to define the language of the app.

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can simply provide access to their workflow in an R script. For instance, this file can be stored on GitHub along with all intermediate files that are created at each given step. Alternatively, given that the information in databases is constantly changing, users can share their R script and associated workspace to ensure that the versions of the retrieved files correspond to specific versions of the databases.

As an example, an early version of this manuscript, written in RMARKDOWN (Baumer & Udwin, 2015), is available from https://github.com/cromanpa94/phruta\_ms. This RMARKDOWN will allow users to replicate the analyses presented in the current paper. With the workspace, saved from the RMARKDOWN as

```
```{r}
save.image(file = "phruta_ms.RData")
...
```

, users can choose to use the stored objects instead of performing searches on GenBank and additional taxonomic databases. These two files confer greater reproducibility to the analyses presented in this article.

### 7 | PERFORMANCE

Assessing the performance of PHRUTA is intrinsically challenging. However, it is expected for functions that scrape and curate information stored in GenBank to show a slow performance on search terms (e.g. clades) with extensive genetic sampling. Below, we focus on comparing the distribution of durations for assembling molecular datasets in the analyses presented in Section 3 in the current document. These estimates cover all the steps outlined above, ranging from using the functions *gene.sampling.retrieve()* and *sq.curate()*. In particular, performance analyses focused on defining the number of simultaneous searches conducted in PHRUTA and the number of examined hits analysing each time from GenBank. Time estimates were calculated using the MICROBENCHMARK R package (Mersmann, 2021).

Figure 5 shows a summary of PHRUTA performance under different search parameters for gene.sampling.retrieve() and sq.curate(). Overall, assembling and curating the analyses presented in Section 3 took consistently less than 6 min to complete, regardless of the parameters used to define searches in PHRUTA. As expected, the more simultaneous searches conducted, the larger the decrease in computational time. Similarly, the higher the number of GenBank hits examined per search, the lower the computational time. These differences related to the number of hits examined are inversely proportional to the number of cores used during the search process. Finally, note that, except when using eight simultaneous cores, searches are all completed successfully. When eight cores are used, only searches for small batches of hits (n = 50 or 100) are successful. We used these patterns to define the optimal search parameters (i.e. fastest and successful) shown in this paper and used for default in PHRUTA and salphycon.

### 8 | ETYMOLOGY AND SYMBOLOGY

Fruta is the Spanish word for Fruit. English ph sounds the same as F in Spanish. In PHRUTA, ph is relative to phylogenetics. I pronounce PHRUTA just as fruta in Spanish. The official logo of the R package features a palenquera (photo by Juan G. Gutiérrez Forero), a symbol of resistance and resilience in Black Latin America, in an imponent Atlas-like position (viz. Greek mythology) carrying frutas (fruits).

### 9 | THE 'SALPHYCON' WEB APP

To support the use of PHRUTA, I developed SALPHYCON, a web app that extends the basic functionality of PHRUTA into a graphical user interface (Figure 6). The app written in Shiny, SALPHYCON, allows users to run the fundamental functions in PHRUTA without the need of writing code. The app also extends PHRUTA to Spanish-speaking users by including translations in the same language using the SHI18NY R package (Marin Diaz, 2022).

### 10 | CONCLUSIONS

Both PHRUTA and SALPHYCON were developed to increase access and reproducibility in phylogenetic analyses. These two tools, released under open access code, following best coding practices, and with transparent implementation, extend the functionality of existing libraries in R and additional software that is commonly used in the phylogenetic workflow. Importantly, PHRUTA and SALPHYCON enable the exploration of existing gene regions and sequences deposited in GenBank to be used in phylogenetic analyses. However, we note, that both of these tools are limited in different ways. Current limitations of PHRUTA include (1) the assemblage of phylogenies with sequences at the species level, (2) the inference of single-gene or multi-locus phylogenies instead of phylogenomics, (3) the implementation of PartitionFinder v1 instead of the latest release (Lanfear et al., 2017), (4) intrinsic limitations associated with the quality of data hosted in GenBank, (5) the lack of flexibility for choosing among more commonly used alignment and phylogenetic packages, among others. Future releases intend to extend the toolbox in PHRUTA to overcome these and other limitations and to explore the implementation of utilities for phylogenetic comparative methods within SALPHYCON.

### **AUTHOR CONTRIBUTIONS**

Cristian Román-Palacios wrote the  $\mbox{\tiny R}$  package, shiny app, vignettes, documentation, developed the associated testing framework and wrote the paper.

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The author thanks Heidi E. Steiner for proofreading the vignettes and documentation in PHRUTA in addition to early versions of this manuscript. The author also thanks Juan G. Gutiérrez Forero for kindly

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### CONFLICT OF INTEREST STATEMENT

The author declares no conflict of interest.

### PEER REVIEW

The peer review history for this article is available at https://www.webofscience.com/api/gateway/wos/peer-review/10.1111/2041-210X.14147.

### DATA AVAILABILITY STATEMENT

The PHRUTA R package code is open source and available on GitHub at https://github.com/ropensci/phruta. The version of PHRUTA used to run the analyses presented in this manuscript is available at on Zenodo (Román-Palacios, 2021) at https://doi.org/10.5281/zenodo.7942117. The associated shiny app, salphycon, in its early stages of development is also available from GitHub at https://github.com/cromanpa94/salphycon. An early version of this manuscript, which is fully reproducible, is available at https://github.com/cromanpa94/phruta\_ms.

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