

taxadb: A High-Performance Local Taxonomic Database Interface

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Abstract

A familiar and growing challenge in ecological and evolutionary research is that of reconciling scientific names of relevant taxa when combining data from separate sources. While this problem is already well understood and numerous naming authorities have been created to address the issue, most researchers lack a fast, consistent, and intuitive way to reconcile taxonomic names. Here, we present **taxadb** R package to address this gap. In contrast to existing tools, **taxadb** provides the following:

- 1) **taxadb** accesses established naming authorities to resolve synonyms, IDs, and hierarchical classification.
- 2) **taxadb** creates a local database, managed automatically from within R, that provides fast operations on millions of taxonomic names.
- 3) **taxadb** provides a consistent and intuitive data format.
- 4) **taxadb** is built on a simple, extensible and language agnostic design that can easily accommodate new authorities.

As ecologists and evolutionary biologists synthesize datasets across larger and larger assemblies of species, we face a continual challenge of reconciling taxonomic names. How many species are in the combined data? Do the studies use the same names for the same species, or do they use different synonyms for the same species? Failing to correct for such differences can lead to significant inflation of species counts and miss-aligned datasets. These challenges have become particularly acute as it becomes increasingly common for researchers to work across a larger number and diversity of species in any given analysis, which may preclude the resources or substantive taxonomic expertise for all clades needed to resolve scientific names (Patterson et al. 2010).

While these issues have long been recognized in the literature (Boyle et al. 2013; Dayrat 2005; Bortolus 2008; Maldonado et al. 2015), and a growing number of databases and tools have emerged over the past few decades (e.g. ITIS 2019; Biotechnology Information 2019; Roskov Y. 2018; Rees 2014; Alvarez and Luebert 2018; Wagner 2016; Foster, Chamberlain, and Grünwald 2018), it remains difficult to resolve taxonomic names to a common authority in a transparent, efficient, and automatable manner. Here, we present an R package, **taxadb**, which seeks to address this gap.

Databases of taxonomic names such as the Integrated Taxonomic Information System (ITIS; ITIS 2019), the National Center for Biological Information's (NCBI) Taxonomy database, (Biotechnology Information 2019), the Catalogue of Life (COL; Roskov Y. 2018), and over one hundred other providers have sought to address these problems by providing expert-curated lists of accepted taxonomic names, synonyms, associated taxonomic rank, hierarchical classification, and scientific authority (e.g. author and date) establishing a scientific name. The R language (R Core Team 2019) is widely used in ecology and evolution (Lai et al. 2019) and the **taxize** package (Chamberlain and Szöcs 2013) has become a popular way for R users to interact with naming providers and name resolution services. **taxize** implements bindings to the web APIs (Application Programming Interface) hosted by many popular taxonomic name providers. Unfortunately, this means that functions in the **taxize** are impacted by several major drawbacks that are inherent in the implementation of these central API servers, such as:

- Queries require internet access at all times.

- Queries are slow and inefficient to implement and perform; frequently requiring separate API calls for each taxonomic name.
- The type of query is highly limited by the API design. For instance, it is usually impossible to make queries across the entire corpus of names, such as “which accepted name has the most known synonyms?”.
- Both query formats and responses differ substantially across different naming providers, making it difficult to apply a script designed for one provider to different provider.
- Most queries are not reproducible, as the results depend on the state of the central server (and potentially the quality of the internet connection). Many names providers update the server data either continuously or at regular intervals, including both revising existing names and adding new names.

Instead of binding existing web APIs, **taxadb** is built around a set of compressed text files which are automatically downloaded, imported, and stored on a local database by **taxadb**. The largest of the taxonomic naming providers today contain under 6 million name records with uncompressed file sizes over a GB, which can be compressed to around 50 MB and downloaded in under a minute on a 1 MB/s connection. By using a local database as the backend, **taxadb** allows R users to interact with large data files without large memory (RAM) requirements. A query for a single name over the web API requires a remote server to respond, execute the query, and serialize the response, which can take several seconds. Thus it does not take many taxa before transferring the entire data set to query locally is more efficient. Moreover, this local copy can be cached on the user’s machine, requiring only the one-time setup, and enabling offline use and reproducible queries. Rather than returning data in whatever format is given by the provider, **taxadb** provides a data structure following a consistent, standardized layout or schema following the Darwin Core Standard (Wieczorek et al. 2012). Table 1 summarizes the list of all naming providers currently accessed by **taxadb**. More details are provided in the Data Sources Vignette, <https://docs.ropensci.org/taxadb/articles/data-sources.html>.

Table 1: Descriptions of the providers supported by **taxadb** with their reference abbreviation and the total number of identifiers contained by each provider.

Provider	Abbreviation	Number of Identifiers	Description
Integrated Taxonomic Information System (ITIS 2019)	itis	597120	originally formed to standardize taxonomic name usage across many agencies in the United States federal government
National Center for Biological Information’s Taxonomy database (NCBI)	ncbi	188221	nomenclature for sequences in the International Nucleotide Sequence Database Collaboration database
Catalogue of Life (COL)	col	1998435	comprehensive taxonomic effort, includes some other providers (e.g. itis)
Global Biodiversity Information Facility Taxonomic Backbone (GBIF 2019)	gbif	3546672	taxonomic backbone of the GBIF database, assembled from other sources including COL
FishBase (Froese and Pauly 2019)	fb	34299	nomenclature for global database of fishes
Open Tree Taxonomy (OTT)	ott	4455820	comprehensive tree of life based on phylogenetic trees and taxonomic data
International Union for Conservation of Nature and Natural Resources (IUCN 2019)	iucn	131927	taxonomy for classification of species status

Package Overview

```
library(tidyverse)
library(taxadb)
```

After loading the package, we look up the taxonomic identifier for Atlantic Cod, *Gadus morhua*, and the compliment:

```
get_ids("Gadus morhua")
```

```
## [1] "ITIS:164712"
```

```
get_names("ITIS:164712")
```

```
## [1] "Gadus morhua"
```

Our first call to any `taxadb` functions will automatically set up a local, persistent database if one has not yet been created. This one-time setup will download, extract, and import the compressed data into persistent database storage (using the appropriate location specified by the operating system (see Ratnakumar, Mick, and Davis 2016), or configured using the environmental variable `TAXADB_HOME`). The example above searches for names in ITIS, the default provider, which can be configured using the `provider` argument. Any future function calls to this function or any other function using data from the same provider will be able to access this data rapidly without the need for processing or an internet connection.

Users can also explicitly trigger this one-time setup using `td_create()` and specifying the provider abbreviation (see Table 1), or simply using `all` to install all available providers:

```
td_create("all")
```

`taxadb` functions like `get_ids()` and `td_create()` take an optional argument, `db`, to an external database connection. `taxadb` will work with most DBI-compliant databases such as MySQL or Postgres, but will be much faster when using a column-oriented database engine such as `duckdb` or `MonetDBLite`. These latter options are also much easier for most users, since each can be installed directly as an R package. `taxadb` will default to the fastest available option. `taxadb` can also run without a database backend by setting `db=NULL`, though some functions will require a lot (2-20 GB) of free RAM for this to work with many of the larger providers.

`taxadb` uses the widely known SQLite database by default, but users are encouraged to install the optional, suggested database backends by passing the option `dependencies = TRUE` to the install command. This installs a `MonetDBLite` database instance (Raasveldt and Mühleisen 2018), a columnar-oriented relational database requiring no additional installation while also providing persistent disk-based storage. This also installs `duckdb`, another local columnar database which is rapidly emerging as an alternative to `MonetDB` and `SQLite`. `taxadb` will automatically detect and use these database engines if available, and automatically handles opening, caching, and closing the database connection. For large queries, `MonetDBLite` or `duckdb` deliver impressive improvements. Our benchmark on resolving the 750 species names in the Breeding Bird Survey against over 3 million names known in the 2019 Catalogue of Life takes 8 minutes in `SQLite` but less than a second in `MonetDBLite`.

Functions in `taxadb` are organized into several families:

- queries that return vectors: `get_ids()` and its complement, `get_names()`,
- queries that filter the underlying taxonomic data frames: `filter_name()`, `filter_rank()`, `filter_id()`, and `filter_common()`,
- database functions `td_create()`, `td_connect()` and `taxa_tbl()`,
- and helper utilities, such as `clean_names()`.

Taxonomic Identifiers

Taxonomic identifiers provide a fundamental abstraction which lies at the heart of managing taxonomic names. For instance, by resolving scientific names to identifiers, we can identify which names are synonyms – different scientific names used to describe the same species – and which names are not recognized. Unmatched names may indicate an error in data entry or otherwise warrant further investigation. Taxon identifiers are also easily resolved to the original authority (scientific publication) establishing the name. (The historical

practice of appending an author and year to a scientific name, e.g. *Poa annua ssp. annua* (Smith 1912), serves a valuable role in disambiguating different uses of the same name but can be notoriously harder to resolve to the appropriate reference, while variation in this convention creates many distinct versions of the same name (Patterson et al. 2010)).

These issues are best illustrated using a minimal example. We'll consider the task of combining data on bird extinction risk as assessed by the IUCN (International Union for Conservation of Nature and Natural Resources 2019) with data on average adult biomass, as estimated in the Elton Traits v1.0 database (Wilman et al. 2016). To keep the example concise enough for visual presentation we will focus on a subset involving just 10 species (Table 2, 3).

```
trait_data <- read_tsv(system.file("extdata", "trait_data.tsv", package="taxadb"))
status_data <- read_tsv(system.file("extdata", "status_data.tsv", package="taxadb"))
```

Table 2: The subset of the IUCN status data used for subsequent taxonomic identifier examples.

iucn_name	category
Pipile pipile	CR
Pipile cumanensis	LC
Pipile cunjubi	LC
Pipile jacutinga	EN
Megapodius decollatus	LC
Scleroptila gutturalis	LC
Margaroperdix madagarensis	LC
Falcipecten falcipecten	NT

Table 3: The subset of the Elton trait data used for subsequent taxonomic identifier examples.

elton_name	mass
Aburria pipile	1816.59
Aburria cumanensis	1239.22
Aburria cunjubi	1195.82
Aburria jacutinga	1240.96
Megapodius reinwardt	666.34
Francolinus leuallantoides	376.69
Margaroperdix madagascariensis	245.00
Catreus wallichii	1436.88
Falcipecten falcipecten	685.61
Falcipecten canadensis	473.65

If we attempted to join these data directly on the species names provided by each table, we would find very little overlap, with only one species having both a body mass and an IUCN threat status resolved (Table 4).

```
joined <- full_join(trait_data, status_data, by = c("elton_name" = "iucn_name"))
```

If we first resolve names used in each data set into shared identifiers, (for instance, using the Catalogue of Life), we discover that there is far more overlap in the species coverage than we might have initially realized.

Table 4: Example IUCN and trait data joined directly on scientific name showing only one match. While common, joining on scientific name does not account for taxonomic inconsistencies between databases and therefore results in seemingly very little overlap in species representation between the two.

elton_name	mass	category
Aburria pipile	1816.59	-
Aburria cumanensis	1239.22	-
Aburria kujubi	1195.82	-
Aburria jacutinga	1240.96	-
Megapodius reinwardt	666.34	-
Francolinus levalliantoides	376.69	-
Margaroperdix madagascariensis	245.00	-
Catreus wallichii	1436.88	-
Falcapennis falcapennis	685.61	NT
Falcapennis canadensis	473.65	-
Pipile pipile	-	CR
Pipile cumanensis	-	LC
Pipile kujubi	-	LC
Pipile jacutinga	-	EN
Megapodius decollatus	-	LC
Scleroptila gutturalis	-	LC
Margaroperdix madagarensis	-	LC

112 First, we just add an ID column to each table by looking up the Catalog of Life identifier for the name
113 provided:

```
traits <- trait_data %>% mutate(id = get_ids(elton_name, "col"))
status <- status_data %>% mutate(id = get_ids(iucn_name, "col"))
```

114 We can now join on the id column instead of names directly:

```
joined <- full_join(traits, status, by = "id")
```

115 This results in many more matches (Table 5), as different scientific names are recognized by the naming
116 provider (Catalog of Life 2018 in this case), as *synonyms* for the same species, and thus resolve to the
117 same taxonomic identifier. While we have focused on a small example for visual clarity here, the `get_ids()`
118 function in `taxadb` can quickly resolve hundreds of thousands of species names to unique identifiers, thanks
119 to the performance of fast joins in a local MonetDBLite database.

Table 5: Example IUCN and trait data joined on taxonomic ID. Multiple species have a different scientific name in the Elton and IUCN Redlist databases but can be match based on their COL taxonomic ID.

elton_name	iucn_name	mass	category	id
Aburria pipile	Pipile pipile	1816.59	CR	COL:35517887
Aburria cumanensis	Pipile cumanensis	1239.22	LC	COL:35537158
Aburria cujubi	Pipile cujubi	1195.82	LC	COL:35537159
Aburria jacutinga	Pipile jacutinga	1240.96	EN	COL:35517886
Megapodius reinwardt	-	666.34	-	COL:35521309
Francolinus levalliantoides	-	376.69	-	COL:35518087
Margaroperdix madagascariensis	Margaroperdix madagarensis	245.00	LC	COL:35521355
Catreus wallichii	-	1436.88	-	COL:35518185
Falciipennis falciipennis	Falciipennis falciipennis	685.61	NT	COL:35521380
Falciipennis canadensis	-	473.65	-	COL:35521381
-	Megapodius decollatus	-	LC	COL:35537166
-	Scleroptila gutturalis	-	LC	-

Box 1: Taxonomic Identifiers and Synonyms

`get_ids()` returns the `acceptedNameUsageID`, the identifier associated with the *accepted* name. Some naming providers, such as ITIS and NCBI, provide taxonomic identifiers to both synonyms and accepted names. Other providers, such as COL and GBIF, only provide identifiers for accepted names. Common practice in Darwin Core archives is to provide an `acceptedNameUsageID` only for names which are synonyms, and otherwise to provide a `taxonID`. For accepted names, the `acceptedNameUsageID` is then given as missing (NA), while for synonyms, the `taxonID` may be missing (NA). In contrast, `taxadb` lists the `acceptedNameUsageID` for accepted names (where it matches the `taxonID`), as well as known synonyms. This is semantically identical, but also more convenient for database interfaces, since it allows a name to mapped to it's accepted identifier (or an identifier to map to it's accepted name usage) without the additional logic. For consistency, we will use the term "identifier" to mean the `acceptedNameUsageID` rather than the more ambiguous `taxonID` (which is undefined for synonyms listed by many providers), unless explicitly stated otherwise.

Unresolved names

`get_ids` offers a first pass at matching scientific names to id, but names may remain unresolved for a number of reasons. First, a name may match to multiple accepted names, as in the case of a species that has been split. By design, these cases are left to be resolved by the researcher using the `filter_` functions to filter underlying taxonomic tables for additional information. A name may also be unresolved due to typos or improper formatting. `clean_names` addresses common formatting issues such as the inclusion of missing species epithets (e.g. `Accipiter sp.`) that prevent matches to the Genus, or intraspecific epithets such as `Colaptes auratus cafer` that prevent matches to the binomial name. These modifications are not appropriate in all settings and should be used with care. Spell check of input names is outside the scope of `taxadb`.

Names may also have an ambiguous resolution wherein a name may be resolved by a different provider than the one specified, either as an accepted name or a synonym. Mapping between providers represent a meaningful scientific statement requiring an understanding of the underlying taxonomic concepts of each provider (Franz and Peet 2009; Franz and Sterner 2018). The spirit of `taxadb` is not to automate steps that require expert knowledge, but provide access to multiple potential "taxonomic theories".

136 *filter_ functions for access to underlying tables*

137 Underlying data tables can be accessed through the family of `filter_` functions, which filter by certain
138 attributes such as scientific name, id, common name, and rank. These functions allow us to ask general
139 questions such as, how many bird species are there?

```
filter_rank("Aves", rank="class", provider = "col") %>%  
  pull(taxonID) %>%  
  n_distinct()
```

140 ## [1] 35398

141 We can also use this to gain a detailed look at specific species or ids. For example, we can explore why
142 `get_ids` fails to resolve a seemingly common species:

```
multi_match <- filter_name("Zosterops rendovae", provider = "col")
```

Table 6: Some names may not resolve to an identifier using `get_ids()` because they match to more than one accepted ID. In such cases `filter_` functions give further detail, as in the example of *Zosterops rendovae* below which has two accepted ID matches.

sort	taxonID	scientificName	acceptedNameUsageID	taxonomicStatus
1	COL:35528603	Zosterops rendovae	COL:35540940	ambiguous synonym
1	COL:35528604	Zosterops rendovae	COL:35542186	ambiguous synonym

143 We see that *Zosterops rendovae* matches two accepted names which the user will have to choose between
144 (Table 6). This is an example of how `taxadb` seeks to provide users with information from existing authorities
145 and names providers, rather than make a potentially arbitrary decision. Because they return `data.frames`,
146 `filter_` functions provide both potential matches. Note that the simpler `get_` functions (`get_ids()`)
147 consider multiple name matches as NA for the id, making them suitable for automated pipelines where
148 manual resolution of duplicates is not an option.

149 *Direct database access*

150 The full taxonomic record in the database can also be directly accessed by `taxa_tbl()`, allowing for
151 whole-database queries that are not possible through the API or web interface of many providers. For
152 example, we can easily check the coverage of accepted species names in each of the classes of vertebrates
153 within the Catalogue of Life (Table 7):

```
verts <- taxa_tbl("col") %>%  
  filter(taxonomicStatus == "accepted", phylum == "Chordata") %>%  
  count(class, sort = TRUE)
```

Box 2: Common Names

`taxadb` can also resolve common names to their identifier by mapping common name to the accepted scientific name. Common names come with many of the same problems as scientific names but often more frequent, e.g. matching to more than one accepted name, non-standardized formatting (like capitalization) and spelling etc. Common names are accessed via `filter_common` which takes a vector of common names. The user can then resolve discrepancies.

154

Table 7: `taxadb` also provides direct access to the database, allowing `dplyr` or SQL queries which can compute across the entire dataset, such as counting accepted species in all vertebrate classes shown here. This kind of query is effectively impossible in most REST API-based interfaces.

class	n
Actinopterygii	32529
Aves	31554
Reptilia	13791
Mammalia	12113
Amphibia	6518
Ascidiacea	2931
Elasmobranchii	1223
Thaliacea	83
Myxini	81
Appendicularia	77
Holocephali	56
Cephalaspidomorphi	45
Leptocardii	30
Sarcopterygii	8

Discussion

Some taxonomic name providers (e.g. OTT, COL) offer periodic releases of a static names list, while many other providers (e.g. ITIS, NCBI, FB, IUCN) offer name data on a rolling basis (i.e. the data returned by a given download URL is updated continuously or at arbitrary intervals without any additional indication if and how that data has changed.) `taxadb`'s `td_create()` function downloads and stores cached snapshots from each provider, which follow an annual release model to support reproducible analyses. All `taxadb` functions that download or access data include an optional argument `version` to indicate which version of the provider data should be used. By default, `taxadb` will determine the latest version available (at the time of writing this is version 2019). Appropriate metadata is stored with each snapshot, including scripts used to access and reformat the data files, as described in the “Data Sources” vignette, <https://docs.ropensci.org/taxadb/articles/data-sources.html>.

Taxonomic identifiers are a powerful tool for maintaining taxonomic consistency, an essential task for a wide variety of applications. Despite multiple taxonomic standardization efforts, resolving names to taxonomic identifiers is often not a standard step in the research work flow due to difficulty in accessing providers and the time consuming API queries necessary for resolving even moderately sized data sets. `taxadb` fills an important gap between existing tools and typical research patterns by providing a fast, reproducible approach for matching names to taxonomic identifiers. `taxadb` is not intended as an improvement or replacement for any existing approaches to taxonomic name resolution. In particular, `taxadb` is not a replacement for the APIs or databases provided, but merely an interface to taxonomic naming information contained within that data.

Lastly, we note that local database design used in `taxadb` is not unique to taxonomic names. Despite the rapid expansion of REST API-based interfaces to ecological data (Boettiger et al. 2015), in our experience, much of the data relevant to ecologists and evolutionary biologists today would be also be amenable to the local database design. The local database approach is much easier for data providers (who can leverage static scientific database repositories instead of maintaining REST servers) and often much faster for data consumers.

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