# 4. Data Workflow

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This tutorial is available as a .qmd on Github.

## **Motivation**

- Familiarize ourselves with more data manipulation tools
- Run through a complete workflow including: database connection, data discovery, data pulling, and data manipulation

## R Data Workflow

Let's run through a realistic data scenario from the beginning.

## Setup

These setup steps will all be familiar to you by now.

```
# minimal packages for RIBBiTR DB data discovery
librarian::shelf(tidyverse, dbplyr, RPostgres, DBI, RIBBiTR-BII/ribbitrrr)

# establish database connection
dbcon = hopToDB("ribbitr")
```

Connecting to database... Success!

```
# load table metadata
mdt = tbl(dbcon, Id("public", "all_tables")) %>%
  filter(table_schema == "survey_data") %>%
  collect()

# load column metadata
mdc = tbl(dbcon, Id("survey_data", "metadata_columns")) %>%
  filter(table_schema == "survey_data") %>%
  collect()
```

## Data discovery and pulling

Suppose we are interested in capture and Bd swab data. Specifically, we want to compare Bd qPCR results for juvenile-to-adult individuals, captured in 2015 or later, across species and sites.

Looking at the table metadata, the two observation tables with most of the data of interest are called "capture" and "bd\_qpcr\_results".

## Pull capture table

### Join with support tables, filter by date

```
# create chain object
chain_capture = tbl_chain("capture", mdc)

# join recursively, filter by date
db_capture_chain = tbl_join(dbcon, chain_capture, tbl = db_capture) %>%
    filter(date >= "2015-01-01")

Joining with bd_swab_lookup ... done.
Joining with survey ... done.
Joining with visit ... done.
Joining with site ... done.
Joining with region ... done.
Joining with country ... done.
```

#### Join with Bd table

### Explore # of samples by life stage, then filter

```
db_capture_bd %>%
  select(life_stage) %>%
  group_by(life_stage) %>%
  summarise(row_count = n()) %>%
  arrange(desc(row_count)) %>%
  collect()
```

```
5 subadult
                                  984
 6 aquatic_larvae
                                  573
 7 metamorph
                                  428
8 larva
                                  417
9 unknown
                                  367
10 terrestrial_development
                                  330
                                  218
11 larvae
12 eggmass
                                    7
13 Unknown
                                    2
db_capture_bd_life = db_capture_bd %>%
  filter(life_stage %in% c("juvenile",
                            "subadult",
                            "adult"),
         !is.na(life_stage))
```

#### Explore # of samples by species, then filter

```
(spp_summary = db_capture_bd_life%>%
  select(species_capture) %>%
  group_by(species_capture) %>%
  summarise(sample_count = n()) %>%
  arrange(desc(sample_count)) %>%
  collect())
```

```
# A tibble: 131 x 2
  species_capture
                         sample count
  <chr>
                                  <int64>
1 rana muscosa
                                    13328
2 rana_clamitans
                                     2366
3 rana_catesbeiana
                                     1601
4 pseudacris_crucifer
                                     1118
5 lithobates_sphenocephalus
                                      965
6 rana_pipiens
                                      551
7 notophthalmus_viridescens
                                      535
8 lithobates_chiricahuensis
                                      453
9 colostethus_panamensis
                                      429
10 hyla_versicolor
                                      417
# i 121 more rows
```

```
(spp_list = spp_summary %>%
  filter(sample_count >= 100) %>%
  pull(species_capture))
```

```
[1] "rana_muscosa"
                                  "rana_clamitans"
 [3] "rana_catesbeiana"
                                  "pseudacris crucifer"
 [5] "lithobates_sphenocephalus" "rana_pipiens"
 [7] "notophthalmus_viridescens" "lithobates_chiricahuensis"
                                  "hyla_versicolor"
 [9] "colostethus_panamensis"
[11] "hyla_chrysoscelis"
                                  "anaxyrus_americanus"
[13] "silverstoneia_flotator"
                                  "lithobates_warszewitschii"
[15] "anaxyrus_fowleri"
                                  "acris_blanchardi"
[17] "rhaebo_haematiticus"
                                  "sachatamia_albomaculata"
[19] "ambystoma_opacum"
                                  "hyla_cinerea"
[21] "lithobates_sylvaticus"
                                  "pseudacris_feriarum"
[23] "smilisca_sila"
                                  "plethodon_glutinosis"
[25] "ambystoma_maculatum"
                                  "plethodon cinereus"
[27] "unknown_species"
                                  "desmognathus_fuscus"
[29] "lithobates_blairi"
db_capture_bd_life_spp = db_capture_bd_life %>%
  filter(species_capture %in% spp_list,
         !is.na(species_capture))
collect data
data_capture_bd_query = db_capture_bd_life_spp %>%
  collect()
colnames(data_capture_bd_query)
 [1] "species_capture"
                             "body_temp_c"
                                                     "life_stage"
 [4] "sex"
                             "capture_animal_state" "bd_swab_id"
 [7] "survey_id"
                             "detection_type"
                                                     "visit_id"
                                                     "site_id"
[10] "date"
                             "survey_time"
[13] "site"
                             "region id"
                                                     "region"
[16] "country_id"
                             "country_name"
                                                     "detected"
[19] "average_target_quant"
head(data_capture_bd_query)
# A tibble: 6 x 19
  species_capture
                     body_temp_c life_stage sex
                                                   capture_animal_state bd_swab_id
  <chr>
                           <dbl> <chr>
                                             <chr> <chr>
                                                                         <chr>>
1 rhaebo_haematiti~
                            NA
                                 juvenile
                                            unkn~ alive
                                                                         150608_04
2 rhaebo_haematiti~
                            NA
                                 juvenile
                                            unkn~ alive
                                                                         150608_05
```

```
3 colostethus_pana~
                        NA adult
                                        unkn~ alive
                                                                    150610_01
4 rhaebo_haematiti~
                          26.7 adult
                                        unkn~ alive
                                                                    150612_01
5 rhaebo_haematiti~
                          25.7 juvenile unkn~ alive
                                                                    150612_02
6 rhaebo_haematiti~
                          26.6 adult
                                         unkn~ alive
                                                                    150612_03
# i 13 more variables: survey_id <chr>, detection_type <chr>, visit_id <chr>,
   date <date>, survey_time <chr>, site_id <chr>, site <chr>, region_id <chr>,
   region <chr>, country_id <chr>, country_name <chr>, detected <dbl>,
    average_target_quant <dbl>
```

These data are ready to be analyzed and visualized!

#### Disconnect

```
dbDisconnect(dbcon)
```

## **Python Data Workflow**

Let's run through a realistic data scenario from the beginning.

## Setup

These setup steps will all be familiar to you by now.

```
# minimal packages for RIBBiTR DB Workflow
import ibis
from ibis import _
import pandas as pd
import dbconfig
import db_access as db

# establish database connection
dbcon = ibis.postgres.connect(**dbconfig.ribbitr)

# load table metadata
mdt = dbcon.table(database = "public", name = "all_tables").to_pandas()

# load column metadata
mdc = (
    dbcon.table(database="public", name="all_columns")
    .filter(_.table_schema == 'survey_data')
```

```
.to_pandas()
)
```

### Data discovery and pulling

Suppose we are interested in capture and Bd swab data. Specifically, we want to compare Bd qPCR results for juvenile-to-adult individuals, captured in 2015 or later, across species and sites.

Looking at the table metadata, the two observation tables with most of the data of interest are called "capture" and "bd\_qpcr\_results".

#### Pull capture table

```
# capture table, select, filter
db_capture = (
  dbcon.table(database="survey_data", name="capture")
  .select([
    'species_capture',
    'body_temp_c',
    'life_stage',
    'sex',
    'capture_animal_state',
    'bd_swab_id',
    'survey_id'
    ])
)
```

### Join with support tables, filter to date

```
# create chain object
chain_capture = db.tbl_chain("capture", mdc)

# join recursively, filter by date
db_capture_chain = (
   db.tbl_join(dbcon, chain_capture, tbl=db_capture)
    .filter(_.date >= '2015-01-01')
   )

Joining with bd_swab_lookup ... done.
Joining with survey ... done.
Joining with visit ... done.
```

```
Joining with site ... done.
Joining with region ... done.
Joining with country ... done.
```

#### Join with Bd table

```
# bd qpcr results lazy table
db_bd_results = (
   dbcon.table(database="survey_data", name="bd_qpcr_results")
   .select([
     'bd_swab_id',
     'detected',
     'average_target_quant'
])
)

# join capture and bd tables
db_capture_bd = (
   db_capture_chain
   .inner_join(db_bd_results, db_capture_chain.bd_swab_id == db_bd_results.bd_swab_id)
)
```

### Explore # of samples by life stage, then filter

```
# count by life stage
life_stage_counts = (
   db_capture_bd
   .group_by('life_stage')
   .aggregate(row_count=_.count())
   .order_by(_.row_count.desc())
   .to_pandas()
   )
print(life_stage_counts)
```

```
life_stage row_count
0
                       adult
                                   22746
1
                        None
                                    4515
2
                    juvenile
                                    3937
3
                     tadpole
                                    1666
4
                    subadult
                                     984
5
              aquatic_larvae
                                     573
6
                   metamorph
                                     428
```

```
7
                       larva
                                    417
8
                    unknown
                                    367
9
    terrestrial_development
                                    330
10
                      larvae
                                    218
                                      7
11
                     eggmass
                                      2
12
                     Unknown
# filter to desired life stages
db_capture_bd_life = (
  db_capture_bd
  .filter(_.life_stage.isin(['juvenile','subadult', 'adult']) & _.life_stage.notnull())
```

## Explore # of samples by species, then filter

```
# count by species
spp_summary = (
   db_capture_bd_life
   .group_by('species_capture')
   .aggregate(sample_count=_.count())
   .order_by(_.sample_count.desc())
   .to_pandas()
   )
print(spp_summary)
```

```
species_capture sample_count
0
                     rana_muscosa
                                            13328
1
                   rana_clamitans
                                             2366
2
                 rana_catesbeiana
                                             1601
3
              pseudacris_crucifer
                                             1118
4
        lithobates_sphenocephalus
                                              965
                silverstoneia_spp
126
                                                1
127
                    diasporus_spp
                                                1
128
           craugastor_monnichorum
                                                1
129
                  acris_crepitans
                                                1
130
     hyalinobatrachium_talamancae
                                                1
```

[131 rows x 2 columns]

```
# generate species list
spp_list = spp_summary[spp_summary['sample_count'] >= 100]['species_capture'].tolist()
```

```
# filter to species in spp_list
db_capture_bd_life_spp = (
   db_capture_bd_life
   .filter(_.species_capture.isin(spp_list) & _.species_capture.notnull())
   )
```

#### Pull data

```
species_capture body_temp_c ... detected average_target_quant
                                    NaN ...
0
        rhaebo_haematiticus
                                                                     0.00
                                                  0.0
                                                                  1007.69
1 lithobates_warszewitschii
                                   18.6 ...
                                                  1.0
                                                                   257.43
2 lithobates_warszewitschii
                                    NaN ...
                                                  1.0
3 lithobates_warszewitschii
                                   21.8 ...
                                                                    56.15
                                                  1.0
4 lithobates_warszewitschii
                                                                     0.00
                                   21.6 ...
                                                  0.0
```

[5 rows x 25 columns]

These data are ready to be analyzed and visualized!

## **Disconnect**

```
# close connection
dbcon.disconnect()
```

<- 3. Data Pulling