

1 Installation

`rethomics` is still under heavy development, so it cannot be uploaded to the CRAN (Comprehensive R Archive Network), where most stable packages are. Instead, we can install R using Hadley Wickham's popular `devtools`¹. Once you have `devtools` installed, it should be straightforward to install `rethomics`

```
> library(devtools)
> install_github("gilestrolab/rethomics", subdir = "rethomics")
```

Check for error messages. Then ensure the package is installed by loading it

```
> library(rethomics)
```

2 Loading data

2.1 Data structure

In `rethomics` the goal is to store *all* the behavioural data in one single dataframe, which is standard for subsequent statistical analysis. In such a dataframe, every row corresponds to a single measurement; that is the position of one animal at one time. Every column describes a statistical variable such as `t`, and `X` and `Y` positions, but they can also hold information about arbitrary conditions such as treatment, sex, age, genotype and so on.

By convention:

- `t`, the variable holding the time, is always *in seconds*
- `X` and `Y`, are *relative to the width of the region* they originate from, and origin is top-left

For a single **experiment**, and when tracking a single animal (i.e. in a single **region**), your data table could look like:

```
> library(rethomics)
> data(multiple_iterative_y_mazes)
> single_animal <- multiple_iterative_y_mazes[experiment_id == "female_FALSE_11.db"]
> print(single_animal)
```

	experiment_id	region_id	sex	sleep_deprived	id	path
1:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
2:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
3:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
4:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
5:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db

¹installation instructions are available here <https://github.com/hadley/devtools/blob/master/R/run-examples.r>

```

4167: female_FALSE_11.db          1 female          FALSE 11 female_FALSE_11.db
4168: female_FALSE_11.db          1 female          FALSE 11 female_FALSE_11.db
4169: female_FALSE_11.db          1 female          FALSE 11 female_FALSE_11.db
4170: female_FALSE_11.db          1 female          FALSE 11 female_FALSE_11.db
4171: female_FALSE_11.db          1 female          FALSE 11 female_FALSE_11.db

```

	t	x	y	w	h	phi	has_interacted
1:	12.166	0.5505208	0.565104167	0.005729167	0.001562500	0	FALSE
2:	12.200	0.5510417	0.564062500	0.008854167	0.003645833	0	FALSE
3:	12.233	0.5520833	0.563020833	0.009375000	0.005729167	0	FALSE
4:	12.266	0.5515625	0.562500000	0.011458333	0.006770833	0	FALSE
5:	12.300	0.5510417	0.561979167	0.011458333	0.007812500	0	FALSE

4167:	151.633	0.6270833	0.002604167	0.009375000	0.003645833	0	FALSE
4168:	151.666	0.6265625	0.002604167	0.007812500	0.003645833	0	FALSE
4169:	151.833	0.6369792	0.001562500	0.003645833	0.001562500	0	FALSE
4170:	151.966	0.6328125	0.001562500	0.004687500	0.001562500	0	FALSE
4171:	152.100	0.6125000	0.003125000	0.004687500	0.004687500	90	FALSE

The first thing you may notice is that the first few columns seem unnecessary as they have constant values. This is because they describe variables that vary between individuals, and we have only one individual.

In order to understand the need for additional columns, we can load data from multiple experiments:

```

> library(rethomics)
> data(multiple_iterative_y_mazes)
> print(multiple_iterative_y_mazes)

```

	experiment_id	region_id	sex	sleep_deprived	id	path
1:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
2:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
3:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
4:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
5:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db

40426:	male_FALSE_25.db	1	male	FALSE	25	male_FALSE_25.db
40427:	male_FALSE_25.db	1	male	FALSE	25	male_FALSE_25.db
40428:	male_FALSE_25.db	1	male	FALSE	25	male_FALSE_25.db
40429:	male_FALSE_25.db	1	male	FALSE	25	male_FALSE_25.db
40430:	male_FALSE_25.db	1	male	FALSE	25	male_FALSE_25.db

	t	x	y	w	h	phi	has_interacted
1:	12.166	0.5505208	0.565104167	0.005729167	0.001562500	0	FALSE
2:	12.200	0.5510417	0.564062500	0.008854167	0.003645833	0	FALSE
3:	12.233	0.5520833	0.563020833	0.009375000	0.005729167	0	FALSE
4:	12.266	0.5515625	0.562500000	0.011458333	0.006770833	0	FALSE
5:	12.300	0.5510417	0.561979167	0.011458333	0.007812500	0	FALSE

```

---
40426: 191.966 0.3661458 0.005208333 0.025520833 0.011458333 36 FALSE
40427: 192.000 0.3651042 0.004687500 0.024479167 0.011458333 37 FALSE
40428: 192.033 0.3645833 0.004166667 0.023958333 0.011979167 35 FALSE
40429: 192.066 0.3625000 0.006770833 0.021875000 0.011979167 0 FALSE
40430: 192.100 0.3625000 0.006250000 0.021875000 0.010937500 0 FALSE

```

The data will *always* have two columns: **experiment_id** and **region_id**. Together, these columns form a *key*, that is combinations of experiment and region represent unique animals. In other words, we can identify, unambiguously, any animal from its region and experiment identifier. Instead of using the legacy `data.frames`, **rethomic** takes advantage of Matt Doyle's powerful `data.table` package². This makes it very easy and efficient to work with large amount of behavioural data. Common operation could involve filtering data and computing variable per condition or per individual. Let us go through several examples:

```

> library(rethomics)
> data(multiple_iterative_y_mazes)
> print(multiple_iterative_y_mazes)

```

	experiment_id	region_id	sex	sleep_deprived	id	path
1:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
2:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
3:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
4:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db
5:	female_FALSE_11.db	1	female	FALSE	11	female_FALSE_11.db

```

---
40426:  male_FALSE_25.db          1  male          FALSE 25  male_FALSE_25.db
40427:  male_FALSE_25.db          1  male          FALSE 25  male_FALSE_25.db
40428:  male_FALSE_25.db          1  male          FALSE 25  male_FALSE_25.db
40429:  male_FALSE_25.db          1  male          FALSE 25  male_FALSE_25.db
40430:  male_FALSE_25.db          1  male          FALSE 25  male_FALSE_25.db

```

	t	x	y	w	h	phi	has_interacted
1:	12.166	0.5505208	0.565104167	0.005729167	0.001562500	0	FALSE
2:	12.200	0.5510417	0.564062500	0.008854167	0.003645833	0	FALSE
3:	12.233	0.5520833	0.563020833	0.009375000	0.005729167	0	FALSE
4:	12.266	0.5515625	0.562500000	0.011458333	0.006770833	0	FALSE
5:	12.300	0.5510417	0.561979167	0.011458333	0.007812500	0	FALSE

```

---
40426: 191.966 0.3661458 0.005208333 0.025520833 0.011458333 36 FALSE
40427: 192.000 0.3651042 0.004687500 0.024479167 0.011458333 37 FALSE
40428: 192.033 0.3645833 0.004166667 0.023958333 0.011979167 35 FALSE
40429: 192.066 0.3625000 0.006770833 0.021875000 0.011979167 0 FALSE
40430: 192.100 0.3625000 0.006250000 0.021875000 0.010937500 0 FALSE

```

²tutorial available at [todo](#)

```

> #WE can simply call this data table `dt'
> dt <- multiple_iterative_y_mazes
> # keeping only females
> dt_female <- dt[sex == 'female',]
> # excluding any data point before 30 secondes (i.e. keeping >= 30s)
> dt_currated <- dt[t >= 30,]
> # Computing, per animal, the time spent in the experiment
> summary_dt = dt[,.(time_spent = max(t) - min(t)),
+                 by=key(dt)]
> print(summary_dt)

```

	experiment_id	region_id	time_spent
1:	female_FALSE_11.db	1	139.934
2:	female_FALSE_2.db	1	149.300
3:	female_TRUE_11.db	1	235.200
4:	female_TRUE_21.db	1	147.866
5:	female_TRUE_4.db	1	181.000
6:	female_TRUE_5.db	1	184.567
7:	female_TRUE_7.db	1	231.133
8:	male_FALSE_25.db	1	184.634

Much more can be achieved using `data.table`, so I would strongly recommend to, at least, read the package introduction³.

2.2 Loading one file

2.3 Fancier queries

2.4 Loading from network drive

2.5 Analysing data

2.6 Sleep annotation

2.7 Visualisation

2.7.1 Overview plot

2.7.2 Ethograms

2.8 Bout analysis