# Package 'rethomics'

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```
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boutAnalysis

Finds 'bouts' in categorical time series.

# Description

This function is used to find contiguous regions of unique value in a – potentially irregular – univariate time series.

# Usage

boutAnalysis(var, data)

# **Arguments**

var the column variable to use in data

data a data.table

# Value

A data.table with columns for the unique value of the bout variable, bout start time, and bout length (ie. duration). Bout analysis will be performed by individual (data.table key), which adds additional columns. Their is one row for each bout.

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#### See Also

rle to perform a run length transform manually

#### **Examples**

buildEthoscopeQuery

Build a query for loading ethoscope dat; using the date of experiments and devices name to retreive result files.

#### **Description**

This function is designed to list and select experimental files. In general, end-users will want to retrieve path to their experimental files according to the date and ID of the video monitor without having to understand the underlying directory structure.

## Usage

```
buildEthoscopeQuery(result_dir, query = NULL, use_cached = FALSE)
```

#### **Arguments**

result\_dir The location of the result directory (i.e. the folder containing all the data).

query An optional query formatted as a dataframe (see details).

use\_cached whether cache files should be used

## **Details**

The optional argument query is expected to be a table where every row maps an experiment. In many respects, it is similar to the what argument in loadEthoscopeData. The only difference is that it does not have a path column. Instead, it must contain two columns:

- date The date and time when the experiment started formatted either as 'yyyy-mm-dd' or 'yyyy-mm-dd\_hh:mm:ss'. In the former case, there may be several matching experiments to a single time (starting the same day). When this happens, *only the last* is returned, and a warning message is displayed.
- machine\_name The name of the machine that acquired the data.

The result is meant to be used directly, as the what argument, by loadEthoscopeData (see examples).

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#### Value

The query extended with the requested paths. When query is not specified, the function returns a table with all available files.

#### Note

The ethoscope platform store data in a hard-coded directory structure /root\_dir/machine\_id/machine\_name/datetim where:

- machine\_id Is, in principle, a universally unique identifier of the acquisition device.
- machine\_name, a human friendly name for acquisition device. In practice, this is expected to be unique within laboratory.
- datetime, the date and time of the start of the experiment

# See Also

cacheEthoscopeData to build a cached data directory.

#### **Examples**

```
# Some sample data are store in the package:
sample_dir <- getSampleDataPath(get_sample_dir = TRUE)
result_dir <- paste(sample_dir,"ethoscope",sep="/")

query <- data.table(
    machine_name = c("E_029","E_014", "E_014"),
    date = c("2016-01-25","2016-01-25", "2016-02-17")
    )
query_path <- buildEthoscopeQuery(result_dir, query)
# now we have maped a query to a path:
print(query)</pre>
```

cacheEthoscopeData

Caches incrementally db files to R native files

# Description

This function is meant to be run by heavy users who want to speed up reading ethoscope file. It will essentially preload all data in a result directory into a cached directory with the same structure. .rdb and .idx fils are generated instead of .db file. In practice, this function will be run every day to fetch and cache new data.

#### Usage

```
cacheEthoscopeData(result_dir, cached_dir, dry_run = F)
```

## **Arguments**

result\_dir The location of the result directory (i.e. the folder containing all the data).

cached\_dir The location of the directory where data should be saved.

dry\_run The location of the directory where data should be saved.

checkColumns 5

checkColumns	Checks if the expected columns are all in a given character vector.
	Through error if not

# Description

Checks if the expected columns are all in a given character vector. Through error if not

# Usage

```
checkColumns(expected_colnames, cols)
```

# Arguments

expected\_colnames

the colnmaes that should be in a dt

cols the actual column names

curateDeadAnimals

Finds when an animal is 'dead' and removes the all consecutive data

# **Description**

In this context, death is defined by very long periods of immobility.

#### Usage

```
curateDeadAnimals(data, max_immobile_live = hours(12))
```

#### **Arguments**

data

the data (i.e a data.table) from a *single* region. It must contain, at least, the columns tand moving.

max\_immobile\_live

the longest duration an alive animal can remain immobile before being considered dead.

#### Value

A data table similar to data where late time points have potentially been removed

#### Note

Death is assumed to be irreversible. Therefore, if an animal is classified as dead, all subsequent data is is removed.

#### See Also

 ${\tt sleepAnnotation} \ and \ {\tt sleepDAMAnnotation} \ to \ define \ movement \ and \ add \ a \ {\tt moving} \ column.$ 

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## **Examples**

```
# Let us load some sample data
data(dam_data)
dt <- dam_data[,</pre>
            sleepDAMAnnotation(.SD),
            by=key(dam_data)]
# let us have a look at the pattern of movement.
# Some animals (e.g. 06, 21, 24) died early.
overviewPlot(moving,dt,normalise_var_per_id = FALSE)
dt_curated <- dt[,curateDeadAnimals(.SD,hours(15)),by=key(dt)]</pre>
# Note that some data has been removed.
# Also, no data was there for region_id == 06, therefore, it is removed altogether
overviewPlot(moving, dt_curated, normalise_var_per_id = FALSE)
# A simple way to compute total lifespan of each remaining animal:
lifespan_dt <- dt_curated[,</pre>
        .(lifespan = max(t) - min(t))
        ,by=key(dt_curated)]
```

dam\_data

A simple rethomics data imported form a Trikinetics DAM2 file.

# **Description**

A 32 animal, 5 days DAM2, single monitor data set formatted in a conventinal rethomics data.table (i.e. has experiment\_id, region\_id and t columns). A dummy variable, condition, was also added for illustration purposes.

## Usage

dam\_data

#### **Format**

a data. table where each row is a single observation of one unique animal at one unique time

#### Author(s)

Anne Petzold, 2015-07-02

# Source

Gilestro lab, Imperial college, London

days 7

Trivially converts days to seconds

# **Description**

Trivially converts days to seconds

#### Usage

days(x)

# Arguments

x number of days

#### Value

the corresponding number of seconds

#### See Also

hours mins

ethogramPlot	Displays the temporal and inter-individual average of a variable of
	interest.

# **Description**

This function produces a graph where the variable of interest and time are on the y and x axes, respectively. It can be used to visualise temporal trends per groups of conditions. The response variable, y, is grouped by time windows of defined size.

#### Usage

```
ethogramPlot(y, data, condition = NULL, facet_var = NULL,
   summary_time_window = mins(30), normalise_var_per_id = FALSE,
   error_bar = NULL, time_wrap = NULL, time_unit_conversion = days)
```

# **Arguments**

y The variable of interest.

data The data.table containing the data. It must have a column with the same name

as y.

condition An optional grouping factor to order rows.

facet\_var An optional grouping factor to draw group in each row of a faceted plot

summary\_time\_window

the width (in seconds) of the time window used to draw each "pixel".

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#### **Details**

time\_wrap is typically used to express time relatively to the start of the day. In other words, it can help be used to pull all days together in one representative day. In this case, time\_wrap=hours(24)'.

At the moment, four types of error bars (error\_bar) are supported:

- 'sd' The standard error
- 'sem' The standard error of the mean (i.e.  $\frac{sd}{\sqrt{n}}$ )
- 'gauss\_ci' The gaussian 95% confidence interval (i.e.  $1.96 \cdot \frac{sd}{\sqrt{n}}$ )
- 'boot\_ci' A standard 95% bootstrap resampling confidence interval. This is done over 5000 replicates. This can be quite *slow*, but is often more statistically sound.

#### Value

A ggplot object that can be plotted directly, or modified.

#### See Also

overviewPlot to show per-individual patterns

```
data(sleep_sexual_dimorphism)
my_data <- sleep_sexual_dimorphism</pre>
# Fraction of animal asleep over time:
p <- ethogramPlot(asleep,my_data)</pre>
# We would like to show that per group:
p <- ethogramPlot(asleep,my_data,condition=sex)</pre>
print(p)
# We can also put error bars:
p <- ethogramPlot(asleep,my_data,condition=sex,error_bar="sem")</pre>
print(p)
# we can also use a condition to split data per row (ggplot faceting):
p <- ethogramPlot(asleep,my_data,condition=sex,facet_var=experiment_id,error_bar="sem")</pre>
print(p)
# p is simply a ggplot object, so we can change things:
print(p + labs(title="MY own title"))
# Let us play with several error bars:
p <- ethogramPlot(asleep,my_data,condition=sex,error_bar="sd")</pre>
p <- ethogramPlot(asleep,my_data,condition=sex,error_bar="sem")</pre>
p <- ethogramPlot(asleep,my_data,condition=sex,error_bar="gauss_ci")</pre>
# this one is a bit slow
p <- ethogramPlot(asleep,my_data,condition=sex,error_bar="boot_ci")</pre>
```

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```
data(dam_data)
# Time, on the x axis, in hours via
p <- ethogramPlot(activity,</pre>
             dam_data,
             condition,
             error_bar = "sem",
              time_unit_conversion=hours # this is where you set time in hours
# summarise/wrap data in one day
p <- ethogramPlot(activity,</pre>
             dam_data,
             condition,
              error_bar = "sem",
              time_wrap=days(1) # this argument does the job
              )
р
```

getSampleDataPath

Get the absolute path to a sample file.

# **Description**

This function is only for testing (and trying) purposes. It provides a way to access raw data (e.g. db files and dam text files) contained within this package.

# Usage

```
getSampleDataPath(path = NULL, get_sample_dir = FALSE)
```

#### **Arguments**

path

The relative path and name of the samples to be loaded. When path is NULL, the function returns the list of all available samples.

get\_sample\_dir whether the function return the root directory of the sample data, nistead of sample files.

### See Also

loadEthoscopeData to read raw experimental data.

hours

Trivially converts hours to seconds

# **Description**

Trivially converts hours to seconds

#### Usage

hours(x)

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## **Arguments**

x number of hours

#### Value

the corresponding number of seconds

#### See Also

days mins

loadDailyDAM2Data

Retrieves DAM2 data from daily saved files

# Description

Uses a query mechanism to get data from a DAM2 array. This is useful when data has been saved, by day, in individual files for each monitor.

# Usage

```
loadDailyDAM2Data(result_dir, query, reference_hour = 9, tz = "UTC",
   verbose = TRUE, FUN = NULL, ...)
```

#### **Arguments**

result\_dir the root directory where all daily data are saved query a formatted query used to request data (see detail).

reference\_hour the hour, in the day, to use as t\_0 reference. This should be expressed on Green-

wich Meridian Time.

tz the time zone on which the DAM2 data was saved (e.g. Europe/London ->

British Summer Time)

verbose whether to print progress (a logical).

FUN an optional function to transform the data from each 'region' (i.e. a data.table)

immediately after is has been loaded.

... extra arguments to be passed to FUN

#### **Details**

query must be a data.table. Conceptually, each row of the query describes the conditions in one channel (when region\_id is specified), or in each monitor (when it is not). It should have the following columns:

- machine\_id the name of the machine used (e.g. 'M002').
- start\_date the first day of the requested experiment (e.g. '2014-12-28').
- stop\_date the last day of the requested experiment (e.g. '2014-12-30').
- region\_id the channel (between 1 and 32) in what the animal was in (e.g. '20'). This is an optional column. If not provided, all 32 channels are loaded with the same conditions.
- ... arbitrary columns to associate conditions/treatments/genotypes/... to the previous columns

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#### Value

A data.table where every row is an individual measurement. That is an activity at a unique time (t) in a unique channel (region\_id), and from a unique result date/experiment (experiment\_id). The time is expressed in seconds. For each different combination of start\_date and machine\_id in the query, an individual experiment\_id is generated.

#### Note

the daily data should be saved in a hard-coded directory structure  ${\tt root\_dir/yyyy/mm/mmdd/mmddMxyz.txt}$ , where:

- yyyy Is the year (e.g. 2014)
- mm and dd, the formatted month and day, respectively (e.g. mm=12 and dd=28).
- xyz, the number of the monitor (e.g 003)

#### See Also

loadDAM2Data to load data from a regular DAM2 file

loadDAM2Data

Retrieves DAM2 data from continuous files

# **Description**

Uses a query mechanism to get data from a DAM2 array. This is useful when using the default behaviour of Trikinetics software where data is simply appended to a single long file per monitor.

# Usage

```
loadDAM2Data(query, FUN = NULL, ...)
```

# Arguments

query a formatted query used to request data (see detail).

FUN an optional function to transform the data from each 'region' (i.e. a data.table) immediately after is has been loaded.

... extra arguments to be passed to FUN

#### **Details**

query must be a data.table. Conceptually, each row of the query describes the conditions in one channel (when region\_id is specified), or in each monitor (when it is not). It should have the following columns:

- $\bullet \ \ path \ the \ location \ of \ your \ data \ file \ (e.g. \ `C:/User/me/Desktop/Monitor3.txt').$
- start\_date the first day of the requested experiment (e.g. '2014-12-28').
- stop\_date the last day of the requested experiment (e.g. '2014-12-30').
- region\_id the channel (between 1 and 32) in what the animal was in (e.g. '20'). This is an optional column. If not provided, all 32 channels are loaded with the same conditions.
- ... arbitrary columns to associate conditions/treatments/genotypes/... to the previous columns

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#### Value

A data.table where every row is an individual measurement. That is an activity at a unique time (t) in a unique channel (region\_id), and from a unique result date/experiment (experiment\_id). The time is expressed in seconds. For each different combination of start\_date and file in the query, an individual experiment\_id is generated.

#### See Also

loadSingleDAM2File to load DAM data that is saved by day

# **Examples**

loadEthoscopeData

Read data from a result file.

# Description

This function is used to convert all the information contained in a result file generated by the ethoscope platform <a href="http://gilestrolab.github.io/ethoscope/">http://gilestrolab.github.io/ethoscope/</a> (i.e a .db file) into an R 'data.table'.

# Usage

```
loadEthoscopeData(what, min_time = 0, max_time = Inf,
  reference_hour = NULL, verbose = TRUE, columns = NULL, ncores = 1,
  FUN = NULL, ...)
```

# **Arguments**

what	an object describing which file(s) to load and, optionally, associated variables/conditions (see details).
min_time	exclude data before min_time (in seconds). This time is relative to the start of the experiment.
max_time	exclude data after max_time (in seconds). It is also relative to the start of the experiment.
reference_hour	the hour, in the day, to use as t_0 reference. When unspecified, time will be relative to the start of the experiment.
verbose	whether to print progress (a logical).

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columns	an optionnal vector of columns to be selected from the db file. Time (t) is always implicitely selected.
ncores	the number of cores to use for optionnal parallel processing.
FUN	an optional function to transform the data from each 'region' (i.e. a data.table) immediately after is has been loaded.
	extra arguments to be passed to FUN

#### **Details**

what can be one of two objects:

- A character vector. In which case, it is assumed that each element is the path to a different file to load.
- A dataframe. The dataframe *must* have a column named 'path'. The path basename will be used as a unique identifier for a specific experiment (experiment\_id). Arbitrary column can be added to map experimental conditions to file name. In addition, the dataframe can have a column named region\_id. When defined, only the specified combinations of path and region\_id will be loaded. This allows to map additional conditions (i.e. data frame columns) to specific regions/files. When additional conditions are provided, they will result in creation of custom columns in the output of this function.

#### Value

A data.table where every row is an individual measurement. That is a position at a unique time (t) in a unique region (region\_id), and from a unique result file/experiment (experiment\_id). The time is expressed in seconds. Distance units (e.g. xy position, height/width) are expressed as a fraction of the width of the region they originate from.

### See Also

loadEthoscopeMetaData To display global informations about a specific file.

```
# First of all, let us load files from the data sample included within this package.
# Most likely, you will already have your own data files.
sample_files <- c("misc/db_files/tube_monitor_validation_subset.db",</pre>
                   "misc/db_files/monitor_validation_subset.db")
# Extract the files in your computer
paths <- sapply(sample_files, getSampleDataPath)</pre>
# Now, `paths` is just a vector of file names:
print(paths)
##################
##################
# Case 1: load ALL REGIONS from a SINGLE FILE
validation_data_file <- paths[1]</pre>
# `validation_data_file` is simply the path to the .db file in your computer
dt <- loadEthoscopeData(validation_data_file)</pre>
print(dt)
##############
# Case 2: load ALL REGIONS from MULTIPLE FILES
# we pass all the files we want to load as the `what` argument
dt <- loadEthoscopeData(paths)</pre>
```

```
# Note the column `experiment_id` in dt. It tells us which file/experiment
# each measurement originates from.
print(dt)
#################
# Case 3: load ALL REGIONS from MULTIPLE FILES AND add CONDITIONS
# Let us imagine that each file/experiment
# was acquired under different experimental condition.
# We can encode this information in a 'master-table' (i.e a data.frame)
# in which a column named \code{path} maps experimental condition(s).
# For instance, 2 different treatments:
master_table <- data.frame(path=paths, treatment=c("control", "drug_A"))</pre>
# Let us check our table:
print(master_table)
# The table looks OK, so we load the actual data
dt <- loadEthoscopeData(master_table)</pre>
# Note that `dt` now contains a column for your treatment.
print(colnames(dt))
# This makes it easier to perform things such as average per treatment.
print(dt[,.(mean_x = mean(x)),by="treatment"])
################
# Case 4: load SELECTED REGIONS from MULTIPLE FILE, WITH CONDITIONS
# Sometimes, different regions contain different conditions.
# If the master table has a column named `region_id`,
# only the specified regions will be returned.
# Let us assume that we want to replicate case 3,
# but, now, we load only the first 20 regions.
master_table <- data.table(path=paths,</pre>
                           treatment=c("control", "drug_A"),
                           region_id=rep(1:20,each= 2))
# We could also imagine that every even region contains a male,
# whilst every odd one has a female:
master_table[, sex := ifelse(region_id %% 2, "male", "female" )]
# Note that we have now two conditions.
# Let us check our new table:
print(master_table)
# Then we can load our data:
dt <- loadEthoscopeData(master_table)</pre>
# This is simply a subset of data, so many regions are missing
# lets display the regions we ended up with
print(dt[,.(NA),by=key(dt)])
#####################
# Case 5: Apply ANALYSIS/function whist loading the data.
# You can also apply a function from this package,
# or your own function to the data as it is being loaded.
# For instance, if you wish to peform a `sleep annotation':
dt <- loadEthoscopeData(paths[1], FUN=sleepAnnotation)</pre>
# You could of course combine this with more conditions/region selection.
# For most complicated cases, you would probably have pre-generated the
# master-table (e.g. as a csv file) before analysing the results.
```

loadSingleDAM2File 15

## **Description**

This function is used to obtain metadata – such as 'time and date of the experiment', 'acquisition device', 'version of the software' and such– embedded in a result file generated by the ethoscope platform.

### Usage

loadEthoscopeMetaData(FILE)

#### **Arguments**

FILE the name of the input file.

# Value

A list containing fields for metadata entries

#### See Also

loadEthoscopeData to load raw data.

## **Examples**

```
## Not run:
FILE <- "result.db"
out <- loadEthoscopeMetaData(FILE)
names(out)
## End(Not run)</pre>
```

loadSingleDAM2File

Read a text file formatted as DAM2 into a single data table.

# **Description**

This function is used to load data from DAM2 devices as a data.table.

#### Usage

```
loadSingleDAM2File(FILE, start_date = -Inf, stop_date = +Inf, tz = "",
    verbose = TRUE)
```

# **Arguments**

FILE the name of the input file.

start\_date the starting date formated as "yyyy-mm-dd" or "yyyy-mm-dd\_hh-mm-ss"

stop\_date the last day of the experiment. Same format as start\_date

tz the time zone of the computer saving the file. By default, tz is taken from the

computer running this function

verbose whether to print progress (a logical).

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#### Value

a data table with an activity (number of beam crosses) variable, a region\_id (channel) variable and a posix time stamp.

#### See Also

loadDAM2Data To load data from a query.

#### **Examples**

```
## Not run:
FILE <- "Monitor53.txt"
out <- loadSingleDAM2File(FILE)
#histogram of x marginal distribution
hist(out[roi_id == 1, x], nclass=100)

## End(Not run)
## Not run:
# More realistic example where we have experimental conditions, and
we want to resample data at 1.0Hz.
# First, the conditions:
conditions <- cbind(roi_id=1:32, expand.grid(treatment=c(T,F), genotype=LETTERS[1:4]))
print(conditions)

## End(Not run)</pre>
```

makeLDAnnotation

Put white and black bars under a plot to show Dark and Light phases.

# **Description**

Put white and black bars under a plot to show Dark and Light phases.

# Usage

```
makeLDAnnotation(p1, time_conversion_unit = days, period = hours(24),
  offset = 0, size = 0.02)
```

# Arguments

# Value

A ggplot object that can be plotted directly, or modified.

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#### **Examples**

```
data(sleep_sexual_dimorphism)
my_data <- sleep_sexual_dimorphism
# Fraction of animal asleep over time:
p <- overviewPlot(asleep,my_data,condition=sex)
p <- makeLDAnnotation(p)
print(p)
p <- ethogramPlot(asleep,my_data,condition=sex,error_bar="sem")
p <- makeLDAnnotation(p)
print(p)</pre>
```

maxVelocityClassifier Motion classifier based on maximum velocity.

# **Description**

Defines whether an animal is moving according to its subpixel velocity. It requires a variable named xy\_dist\_log10x1000 in the .db file.

# Usage

```
maxVelocityClassifier(data, velocity_threshold = 0.006)
```

# **Arguments**

```
data the data.table containing behavioural features used for movement classification. velocity_threshold velocity above which an animal is classified as 'moving'.
```

# Value

a data table with the columns moving (logical, TRUE iff. motion was detected) and t\_round (the 'rounded' time). There is one row per rounded time point.

# See Also

sleepAnnotation to apply this function to all subsequent time windows.

mins

Trivially converts minutes to seconds

# Description

Trivially converts minutes to seconds

# Usage

```
mins(x)
```

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#### **Arguments**

Х

number of minutes

# Value

the corresponding number of seconds

#### See Also

days hours

```
multiple_iterative_y_mazes
```

Unprocessed ethoscope data of animals waling in a y maze.

# **Description**

An ,8 animal, single region\_id data set formatted in a conventinal rethomics data.table (i.e. has experiment\_id, region\_id and t columns).

## Usage

```
multiple_iterative_y_mazes
```

#### **Format**

a data. table where each row is a single observation of one unique animal at one unique time

# Author(s)

Diana Bicazan, 2015-08-05

## **Source**

Gilestro lab, Imperial college, London

overviewPlot

Displays, per individual, the temporal average of a variable of interest.

# **Description**

This function produces a tiled representation in which every row represents one individual (i.e. from a unique combination of region and experiment). The x axis represents time in days. The values of the variable of interest are represented by different colour intensities.

# Usage

```
overviewPlot(y, data, condition = NULL, summary_time_window = mins(30),
  normalise_var_per_id = FALSE, time_wrap = NULL,
  time_unit_conversion = days)
```

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#### **Arguments**

```
The variable of interest

data

The data.table containing the data. It must have a column with the same name as y.

condition

An optional grouping factor to order rows.

summary_time_window

the width (in seconds) of the time window used to draw each pixel.

normalise_var_per_id

whether each row is to be normalised, using new_y = (y - mean(y))/sd(y).

time_wrap

the time (in seconds) used to wrap the data (see details).

time_unit_conversion

a function to convert time in the x axis. typically, days, hours or mins.
```

#### Details

time\_wrap is typically used to express time relatively to the start of the day. In other words, it can help be used to pull all days together in one representative day. In this case, time\_wrap=hours(24)'.

#### Value

A ggplot object that can be plotted directly or modified.

#### See Also

ethogramPlot To show trend by aggregating individuals over time.

```
# Load sample data, it is already annotated for sleep, has sex=='male' or sex=="female"
data(sleep_sexual_dimorphism)
my_data <- sleep_sexual_dimorphism</pre>
\# let us have a look of the max velocity as a measure of activity
p <- overviewPlot(max_velocity,my_data)</pre>
print(p)
# what about sleep amount?
p <- overviewPlot(asleep,my_data)</pre>
print(p)
# we can also group by condition. For instance by sex:
p <- overviewPlot(asleep,my_data,condition = sex)</pre>
print(p)
# p is simply a ggplot object, so we can change things:
print(p + labs(title="MY own title"))
##### time wrapping example
data(dam_data)
# the original plot:
p <- overviewPlot(activity,dam_data)</pre>
# summarise/wrap activity in one `day'
p <- overviewPlot(activity,dam_data,time_wrap=hours(24))</pre>
#####expresses time in hours:
p <- overviewPlot(activity,dam_data, time_unit_conversion=hours)</pre>
```

20 sleepAnnotation

sleepAnnotation

Determines whether an animal is asleep

# **Description**

This function uses a motion classifier to first decide whether an animal is moving during a given time window. Then, it defines sleep as contiguous immobility for a minimal duration.

# Usage

```
sleepAnnotation(data, time_window_length = 10, min_time_immobile = 60 * 5,
  motion_classifier_FUN = maxVelocityClassifier, ...)
```

#### **Arguments**

```
data the data (i.e a data.table) from a single region. It must contain, at least, the columns 't', 'x' and 'y'.

time_window_length

The number of seconds to be used by the motion classifier. This corresponds to the sampling period of the output data.

min_time_immobile

the minimal duration (in s) after which an immobile animal is scored as 'asleep'.

motion_classifier_FUN

the function used to classify movement.

... extra arguments to be passed to motion_classifier_FUN
```

# Value

A data table similar to data with additional variables/annotations (i.e. 'moving', 'asleep').

#### Note

The resulting data will only have one data point every time\_window\_length seconds.

#### See Also

loadEthoscopeData to load data and optionally apply analysis on the fly.

```
# Let us load some sample data
data(tube_monitor_validation)
# We will start only with region 2:
dt_region2 <- tube_monitor_validation[region_id==2,]
sleep_dt <- sleepAnnotation(dt_region2)
print(sleep_dt)
# We make a sleep `barecode'
ggplot(sleep_dt, aes(t,region_id,fill=asleep)) + geom_tile()
# A bit of data.table wizardry to apply that to each experiement and region:
sleep_dt <- tube_monitor_validation[,sleepAnnotation(.SD),by=key(tube_monitor_validation)]
# The same bare code for all regions
ggplot(sleep_dt, aes(t,region_id,fill=asleep)) + geom_tile()</pre>
```

sleepDAMAnnotation 21

sleepDAMAnnotation

Determines whether an animal is asleep using beam crossing activity

# Description

Sleep as contiguous inactivity (absence of beam crossing) for a minimal duration.

# Usage

```
sleepDAMAnnotation(data, time_window_length = 60, min_time_immobile = 60 \star 5)
```

# **Arguments**

data

the data (i.e a data.table) from a *single* region. It must contain, at least, the columns t, x and y.

time\_window\_length

The number of seconds to be used by the motion classifier. This corresponds to the sampling period of the output data.

min\_time\_immobile

the minimal duration (in s) after which an immobile animal is scored as 'asleep'.

# Value

A data table similar to data with additional variables/annotations (i.e. 'moving', 'asleep').

#### Note

The resulting data will only have one data point every time\_window\_length seconds.

# See Also

loadDAM2Data To load DAM2 data first/ apply this function to each animal.

sleep\_sexual\_dimorphism

Sample of ethoscope data showing difference between males and females.

# **Description**

A sleep-annotated 56 animal data set.Movement was defined for 10s time windows. conventinal rethomics data.table (i.e. has experiment\_id, region\_id and t columns).

# Usage

sleep\_sexual\_dimorphism

#### **Format**

a data.table where each row is a single observation of one unique animal at one unique time

#### Author(s)

Quentin Geissmann, 2015-06-13

#### **Source**

Gilestro lab, Imperial college, London

```
tube_monitor_validation
```

A dataset to validate tracking algorithms.

# **Description**

A 20 animal single monitor data set formatted in a conventinal rethomics data.table (i.e. has experiment\_id, region\_id and t columns). Only raw tracking values are present in dataset.

# Usage

tube\_monitor\_validation

#### **Format**

a data. table where each row is a single observation of one unique animal at one unique time

## Author(s)

Quentin Geissmann, 2015-05-02

### Source

Gilestro lab, Imperial college, London

virtualBeamCrossClassif 23

virtualBeamCrossClassif

Motion classifier based on beam crosses.

# Description

Defines whether an animal is moving. This is achieved by computing the number of crossed of a "virtual beam" in the middle of its region (i.e. at x=0.5). This emulate the type of data generated by DAM2.

# Usage

virtualBeamCrossClassif(data)

# Arguments

data

the data.table containing behavioural features used for movement classification.

# Value

a data table with the columns moving (logical, TRUE iff. motion was detected) and t\_round (the 'rounded' time). There is one row per rounded time point.

# See Also

maxVelocityClassifier to defince movement by maximum velocity, which is more accurate, instead.

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