# Package 'celltrackR'

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Type Package

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**Description** Provides methodology to analyze cells that move in a two- or three-dimensional space. Available measures include displacement, confinement ratio, autocorrelation, straightness, turning angle, and fractal dimension. Measures can be applied to entire tracks, steps, or subtracks with varying length. While the methodology has been developed for cell trajectory analysis, it is applicable to anything that moves including animals, people, or vehicles. Some of the methodology implemented in this packages was described by: Beauchemin, Dixit, and Perelson (2007) <doi:10.4049/jimmunol.178.9.5505>, Beltman, Maree, and de Boer (2009) <doi:10.1038/nri2638>, Gneiting and Schlather (2004) <doi:10.1137/S0036144501394387>, Mokhtari, Mech, Zitzmann, Hasenberg, Gunzer, and Figge (2013) <doi:10.1371/journal.pone.0080808>, Moreau, Lemaitre, Terriac, Azar, Piel, Lennon-Dumenil, and Bousso (2012) <doi:10.1016/j.immuni.2012.05.014>, Textor, Peixoto, Henrickson, Sinn, von Andrian, and Westermann (2011) <doi:10.1073/pnas.1102288108>, Textor, Sinn, and de Boer (2013) <doi:10.1186/1471-2105-14-S6-S10>, Textor, Henrickson, Mandl, von Andrian, Westermann, de Boer, and Beltman (2014) <doi:10.1371/journal.pcbi.1003752>. **Depends** R (>= 3.4.0)

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# **R** topics documented:

aggregate.tracks	. 3
analyzeCellPairs	. 6
analyzeStepPairs	. 7
AngleAnalysis	. 8
angleCells	. 10
angleSteps	. 10
angleToDir	. 11
angleToPlane	. 12
angleToPoint	. 14
applyStaggered	. 15
as.data.frame.tracks	. 16
as.list.tracks	. 17
as.tracks.data.frame	. 18
BCells	. 19
peaucheminTrack	. 20
pootstrapTrack	. 21
poundingBox	. 22
prownianTrack	. 23
cellPairs	. 24
celltrackR	. 24
cheatsheet	. 26
clusterTracks	. 27
listanceCells	. 28
listanceSteps	. 29
listanceToPlane	. 30
listanceToPoint	. 31
ilterTracks	. 32
getFeatureMatrix	. 32
notellingsTest	. 33
nterpolateTrack	. 34
naxTrackLength	. 35
Neutrophils	. 36

2

normalizeToDuration	37
normalizeTracks	37
plot.tracks	38
plot3d	39
plotTrackMeasures	39
prefixes	41
projectDimensions	42
read.tracks.csv	42
repairGaps	44
selectSteps	45
selectTracks	46
simulateTracks	46
sort.tracks	47
splitTrack	48
staggered	48
stepPairs	49
subsample	50
subtracks	50
subtracksByTime	51
TCells	52
timePoints	53
timeStep	54
trackFeatureMap	55
TrackMeasures	56
tracks	59
vecAngle	60
wrapTrack	61
	62

## Index

aggregate.tracks Compute Summary Statistics of Subtracks

## Description

Computes a given measure on subtracks of a given track set, applies a summary statistic for each subtrack length, and returns the results in a convenient form. This important workhorse function facilitates many common motility analyses such as mean square displacement, turning angle, and autocorrelation plots.

## Usage

```
## S3 method for class 'tracks'
aggregate(
    x,
    measure,
    by = "subtracks",
    FUN = mean,
```

```
subtrack.length = seq(1, (maxTrackLength(x) - 1)),
max.overlap = max(subtrack.length),
na.rm = FALSE,
filter.subtracks = NULL,
count.subtracks = FALSE,
...
```

## Arguments

X	the tracks object whose subtracks are to be considered. If a single track is given, it will be coerced to a tracks object using wrapTrack (but note that this requires an explicit call aggregate.tracks).	
measure	the measure that is to be computed on the subtracks.	
by	a string that indicates how grouping is performed. Currently, two kinds of group- ing are supported:	
	• "subtracks" Apply measure to all subtracks according to the parameters subtrack.length and max.overlap.	
	• "prefixes" Apply measure to all prefixes (i.e., subtracks starting from a track's initial position) according to the parameter subtrack.length.	
FUN	a summary statistic to be computed on the measures of subtracks with the same length. Can be a function or a string. If a string is given, it is first matched to the following builtin values:	
	• "mean.sd" Outputs the mean and $mean - sd$ as lower and $mean + sd$ as upper bound	
	• "mean.se" Outputs the mean and $mean - se$ as lower and $mean + se$ as upper bound	
	• "mean.ci.95" Outputs the mean and upper and lower bound of a parametric 95 percent confidence intervall.	
	• "mean.ci.99" Outputs the mean and upper and lower bound of a parametric 95 percent confidence intervall.	
	• "iqr" Outputs the interquartile range, that is, the median, and the 25-percent- quartile as a lower and and the 75-percent-quartile as an upper bound	
	If the string is not equal to any of these, it is passed on to match. fun.	
subtrack.length		
	an integer or a vector of integers defining which subtrack lengths are considered. In particular, subtrack.length=1 corresponds to a "step-based analysis" (Beltman et al, 2009).	
max.overlap	an integer controlling what to do with overlapping subtracks. A maximum over- lap of max(subtrack.length) will imply that all subtracks are considered. For a maximum overlap of 0, only non-overlapping subtracks are considered. A neg- ative overlap can be used to ensure that only subtracks a certain distance apart are considered. In general, for non-Brownian motion there will be correlations between subsequent steps, such that a negative overlap may be necessary to get a proper error estimate.	

4

na.rm	logical. If TRUE, then NA's and NaN's are removed prior to computing the summary statistic.	
filter.subtrack	(S	
	a function that can be supplied to exclude certain subtracks from an analysis. For instance, one may wish to compute angles only between steps of a certain minimum length (see Examples).	
count.subtracks		
	logical. If TRUE, the returned dataframe contains an extra column ntracks showing the number of subtracks of each length. This is useful to keep track of since the returned value estimates for high i are often based on very few subtracks.	
	further arguments passed to or used by methods.	

#### Details

For every number of segments i in the set defined by subtrack.length, all subtracks of any track in the input tracks object that consist of exactly i segments are considered. The input measure is applied to the subtracks individually, and the statistic is applied to the resulting values.

#### Value

A data frame with one row for every i specified by subtrack.length. The first column contains the values of i and the remaining columns contain the values of the summary statistic of the measure values of tracks having exactly i segments.

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

## Examples

```
## A mean square displacement plot with error bars.
dat <- aggregate(TCells, squareDisplacement, FUN="mean.se")
with( dat ,{
    plot( mean ~ i, xlab="time step",
    ylab="mean square displacement", type="l" )
    segments( i, lower, y1=upper )
} )
## Note that the values at high i are often based on very few subtracks:
msd <- aggregate( TCells, squareDisplacement, count.subtracks = TRUE )
tail( msd )
## Compute a turning angle plot for the B cell data, taking only steps of at least
## 1 micrometer length into account
check <- function(x) all( sapply( list(head(x,2),tail(x,2)), trackLength ) >= 1.0 )
plot( aggregate( BCells, overallAngle, subtrack.length=1:10,
    filter.subtracks=check )[,2], type='l' )
```

```
## Compare 3 different variants of a mean displacement plot
# 1. average over all subtracks
plot( aggregate( TCells, displacement ), type='1' )
# 2. average over all non-overlapping subtracks
lines( aggregate( TCells, displacement, max.overlap=0 ), col=2 )
# 3. average over all subtracks starting at 1st position
lines( aggregate( TCells, displacement, by="prefixes" ), col=3 )
```

analyzeCellPairs Find Distances and Angles for all Pairs of Tracks

### Description

Find all pairs of cells and return the shortest distance between them at any point, as well as the angle between their overall displacement vectors.

## Usage

```
analyzeCellPairs(X, ...)
```

#### Arguments

X	a tracks object
	further arguments passed on to angleCells

#### Details

Analyzing track angles at different distances can be useful to detect directional bias or local crowding effects; see (Beltman et al, 2009).

Internally, the function uses cellPairs, angleCells, and distanceCells.

#### Value

A dataframe with four columns: two for the indices of cellpairs, one for the distance between them, and one for their angle.

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

#### See Also

analyzeStepPairs to do something similar for single steps rather than entire tracks.

6

#### analyzeStepPairs

## Examples

```
## Plot distance versus angle for all cell pairs
pairs <- analyzeCellPairs( TCells )
scatter.smooth( pairs$dist, pairs$angle )</pre>
```

analyzeStepPairs Find Distances and Angles for all Pairs of Steps

#### Description

Find cell indices and timepoints where these cells both have a step, then return angles and distances for each pair of steps.

#### Usage

```
analyzeStepPairs(X, filter.steps = NULL, ...)
```

## Arguments

Х	a tracks object
filter.steps	optional: a function used to filter steps on. See examples.
	further arguments passed on to angleSteps

#### **Details**

Analyzing step angles at different distances can be useful to detect directional bias or local crowding effects; see (Beltman et al, 2009).

Internally, the function uses stepPairs, angleSteps, and distanceSteps.

#### Value

A dataframe with five columns: two for the indices of cellpairs that share a step, one for the timepoint at which they do so, one for the distance between them, and one for their angle.

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

#### See Also

analyzeCellPairs to do something similar for entire tracks rather than single steps.

#### Examples

```
## Plot distance versus angle for all step pairs, filtering for those that
## displace at least 2 microns
pairs <- analyzeStepPairs( TCells, filter.steps = function(t) displacement(t) > 2 )
scatter.smooth( pairs$dist, pairs$angle )
```

AngleAnalysis

## Description

Analyzing angles to reference directions, points, or planes can be useful to detect artefacts and/or directionality in tracking datasets (Beltman et al, 2009). All these functions take a track and a reference (point/direction/plane) as input and return a distance or angle as output. Angles/distances are by default computed to the first step in the given track.

#### Details

angleToPoint and distanceToPoint return the angle/distance of the track to the reference point. The distance returned is between the first coordinate in the track and the reference point. The angle is between the direction of the first step in the track and the line between its first coordinate and the reference point. Angles are by default returned in degrees, use degrees=FALSE to obtain radians. These functions are useful to detect directional bias towards a point of interest, which would result in an average angle of less than 90 degrees with the reference point (especially for tracks at a small distance to the reference point).

angleToPlane and distanceToPlane return the angle/distance of the track to a plane of interest. This plane must be specified by three points lying on it. The distance returned is between the first coordinate in the track and the reference point. The angle is between the direction of the first step in the track and the plane of interest. These functions are useful to detect tracking artefacts near the borders of the imaging volume. Use boundingBox to guess where those borders are. Angles are by default returned in degrees, use degrees=FALSE to obtain radians.

angleToDir returns the angle of the first step in a track to a direction of interest. This function is useful to detect directionality in cases where the direction of the bias is known in advance (e.g. when cells are known to move up a chemotactic gradient): in that case, the average angle to the reference direction should be less than 90 degrees. Angles are by default returned in degrees, use degrees=FALSE to obtain radians.

angleSteps and distanceSteps return the angle/distance between a pair of steps in the data that occur at the same timepoint. Angles are in degrees by default, use degrees=FALSE to obtain radians. Use stepPairs to extract all pairs of steps that occur at the same timepoint, and use analyzeStepPairs to do this and then also obtain the angles and distances for each of these pairs.

angleCells and distanceCells return the angle/distance between a pair of tracks in the data. The computed angles are between the overall displacement vectors of the tracks, the distance is the shortest distance between them at any timepoint they share. Angles are in degrees by default, use degrees=FALSE to obtain radians. Use cellPairs to extract all pairs of cells in the data, and use analyzeCellPairs to do this and then also obtain the angles and distances for each of these pairs.

#### Value

This page is for documentation only and provides an overview of angle analysis functions and their use cases. The return values of each of these functions are documented separately; please follow the link to the documentation page of that specific function.

#### AngleAnalysis

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

#### See Also

TrackMeasures for other measures that can be used to quantify tracks.

See the vignettes on Quality Control and Track Analysis for more detailed examples of angle analyses. browseVignettes(package = "celltrackR")

#### Examples

```
## Plotting the angle versus the distance to a reference point can be informative to
## detect biased movement towards that point. We should be suspicious especially
## when small angles are more frequent at lower distances.
steps <- subtracks( Neutrophils, 1 )
bb <- boundingBox( Neutrophils )
angles <- sapply( steps, angleToPoint, p = bb["max",-1] )
distances <- sapply( steps, distanceToPoint, p = bb["max",-1] )
scatter.smooth( distances, angles )
abline( h = 90, col = "red" )</pre>
```

```
## Get a distribution of Neutrophil step angles with the reference direction
## in positive y direction. The histogram is enriched for low angles, suggesting
## directed movement:
hist( sapply( steps, angleToDir, dvec=c(0,1,0) ) )
```

```
## Plotting the angle versus the distance to a reference plane can be informative to
## detect tracking artefacts near the border of the imaging volume.
## We should be suspicious especially when small angles are more frequent at low distances
## to the border planes.
steps <- subtracks( TCells, 1 )
minz <- boundingBox( TCells )["min","z"]
## Compute angles and distances to the lower plane in z-dimension
angles <- sapply( steps, angleToPlane, p1 = c(0,0,minz), p2 = c(1,0,minz), p3 = c(0,1,minz) )
distances <- sapply( steps, distanceToPlane, p1 = c(0,0,minz), p2 = c(1,0,minz), p3 = c(0,1,minz) )
scatter.smooth( distances, angles )
abline( h = 32.7, col = "red" )
```

```
## Plot distance versus angle for all cell pairs
pairs <- analyzeCellPairs( TCells )
scatter.smooth( pairs$dist, pairs$angle )
abline( h = 90, col = "red" )</pre>
```

```
## Plot distance versus angle for all step pairs, filtering for those that
## displace at least 2 microns
pairs <- analyzeStepPairs( TCells, filter.steps = function(t) displacement(t) > 2 )
scatter.smooth( pairs$dist, pairs$angle )
abline( h = 90, col = "red" )
```

angleCells

#### Description

Compute the angle between the displacement vectors of two tracks in the dataset.

## Usage

angleCells(X, cellids, degrees = TRUE)

## Arguments

Х	a tracks object
cellids	a vector of two indices specifying the tracks to get steps from.
degrees	logical; should angle be returned in degrees instead of radians? (defaults to $\ensuremath{TRUE}\xspace)$

## Value

A single angle.

## See Also

distanceCells to compute the minimum distance between the tracks, and AngleAnalysis for other methods to compute angles and distances.

## Examples

```
## Find the angle between the tracks with ids 1 and 2 angleCells( TCells, c("1", "2") )
```

angleSteps Angle between Two Steps

## Description

Compute the angle between two steps in the dataset that occur at the same timepoint.

#### Usage

```
angleSteps(X, trackids, t, degrees = TRUE, quietly = FALSE)
```

## angleToDir

#### Arguments

Х	a tracks object
trackids	a vector of two indices specifying the tracks to get steps from.
t	the timepoint at which the steps should start.
degrees	logical; should angle be returned in degrees instead of radians? (defaults to TRUE)
quietly	logical; should a warning be returned if one or both of the steps are missing in the data and the function returns NA?

## Value

A single angle, or NA if the desired timepoint is missing for one or both of the tracks.

#### See Also

distanceSteps to compute the distance between the step starting points, timePoints to list all timepoints in a dataset, and AngleAnalysis for other methods to compute angles and distances.

## Examples

```
## Find the angle between the steps of the tracks with ids 1 and 2, at the 3rd
## timepoint in the dataset.
t <- timePoints( TCells )[3]
angleSteps( TCells, c("1","2"), t )</pre>
```

angleToDir Angle with a Reference Direction

## Description

Compute the angle between the first step of a track and a reference direction. Useful to detect biased movement when the directional bias is known (see examples).

## Usage

```
angleToDir(x, dvec = c(1, 1, 1), from = 1, xdiff = diff(x), degrees = TRUE)
```

#### Arguments

x	a single input track; a matrix whose first column is time and whose remaining columns are a spatial coordinate.
dvec	numeric vector specifying a reference direction to compute angles to.
from	index, or vector of indices, of the first row of the track. If from is a vector, angles are returned for all steps starting at the indices in from.
xdiff	row differences of x.
degrees	logical; should angles be returned in degrees rather than radians? (default = TRUE).

#### Details

The average angle of steps to a reference direction should be 90 degrees if there is no bias towards movement in the direction of the reference point. If there is such a bias, there should be an enrichment of smaller angles. The expected distribution without bias is a uniform distribution in 2D or a sine distribution in 3D (Beltman et al, 2009).

#### Value

A single angle.

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

#### See Also

AngleAnalysis for other methods to compute angles and distances.

#### Examples

```
## Get a distribution of Neutrophil step angles with the reference direction in positive
## y direction. The histogram is enriched for low angles, suggesting directed movement:
steps <- subtracks( Neutrophils, 1 )
hist( sapply( steps, angleToDir, dvec=c(0,1,0) ) )
```

angleToPlane

Angle with a Reference Plane

#### Description

Compute the angle between the first step of a track and a reference plane. Useful to detect directed movement and/or tracking artefacts.

## Usage

```
angleToPlane(
    x,
    p1 = c(0, 0, 0),
    p2 = c(0, 1, 0),
    p3 = c(1, 0, 0),
    from = 1,
    xdiff = diff(x),
    degrees = TRUE
)
```

#### angleToPlane

#### Arguments

x	a single input track; a matrix whose first column is time and whose remaining columns are a spatial coordinate.
p1, p2, p3	numeric vectors of coordinates of three points specifying a reference plane to compute distances to.
from	index, or vector of indices, of the first row of the track. If from is a vector, angles are returned for all steps starting at the indices in from.
xdiff	row differences of x.
degrees	logical; should angles be returned in degrees rather than radians? (default = TRUE).

## Details

The average angle of steps to a reference plane should be roughly 32.7 degrees. Lower angles to the border planes of an imaging volume can be indicative of tracking artefacts, and systematic deviations from 32.7 can indicate a directional bias (Beltman et al, 2009).

## Value

A single angle.

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

#### See Also

distanceToPlane to compute the distance to the reference plane, and AngleAnalysis for other methods to compute angles and distances.

#### Examples

```
## Plotting the angle versus the distance to a reference plane can be informative to
## detect tracking artefacts near the border of the imaging volume.
## We should be suspicious especially when small angles are more frequent at low distances
## to the border planes.
steps <- subtracks( TCells, 1 )
minz <- boundingBox( TCells )["min","z"]
## Compute angles and distances to the lower plane in z-dimension
angles <- sapply( steps, angleToPlane, p1 = c(0,0,minz), p2 = c(1,0,minz), p3 = c(0,1,minz) )
distances <- sapply( steps, distanceToPlane, p1 = c(0,0,minz), p2 = c(1,0,minz), p3 = c(0,1,minz) )
scatter.smooth( distances, angles )
abline( h = 32.7, col = "red" )</pre>
```

angleToPoint

## Description

Compute the angle between the first step of a track and a reference point. Useful to detect directed movement towards a point (see examples).

#### Usage

```
angleToPoint(x, p = c(1, 1, 1), from = 1, xdiff = diff(x), degrees = TRUE)
```

#### Arguments

х	a single input track; a matrix whose first column is time and whose remaining columns are a spatial coordinate.
р	numeric vector of coordinates of the reference point p to compute angles/distances to.
from	index, or vector of indices, of the first row of the track. If from is a vector, angles are returned for all steps starting at the indices in from.
xdiff	row differences of x.
degrees	logical; should angles be returned in degrees rather than radians? (default = TRUE).

## Details

The average angle of steps to a reference point should be 90 degrees if there is no bias towards movement in the direction of the reference point. If there is such a bias, there should be an enrichment of smaller angles. The expected distribution without bias is a uniform distribution in 2D or a sine distribution in 3D (Beltman et al, 2009).

#### Value

A single angle.

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

#### See Also

distanceToPoint to compute the distance to the reference point, and AngleAnalysis for other methods to compute angles and distances.

## applyStaggered

#### Examples

```
## Get a distribution of step angles with a reference point
## Use bb to get the corner with highest x,y,and z value
## The histogram is enriched for low angles, suggesting directed movement:
steps <- subtracks( Neutrophils, 1 )
bb <- boundingBox( Neutrophils )
hist( sapply( steps, angleToPoint, p = bb["max",-1] ) )
## The same does not hold for movement of T cells towards the point (0,0,0)
steps <- subtracks( TCells, 1 )
hist( sapply( steps, angleToPoint, p = c(0,0,0) ) )
## Plotting the angle versus the distance to the reference point can also be informative,
## especially when small angles are more frequent at lower distances.
angles <- sapply( steps, angleToPoint, p = bb["max",-1] )
distances <- sapply( steps, distanceToPoint, p = bb["max",-1] )
scatter.smooth( distances, angles )
abline( h = 90, col = "red" )
```

applyStaggered

Compute a Measure on a Track in a Staggered Fashion

#### Description

Computes a measure on all subtracks of a track and return them either as a matrix or return their mean.

### Usage

```
applyStaggered(x, measure, matrix = FALSE, min.segments = 1)
```

#### Arguments

х	the track for which the measure is to be computed.
measure	the measure that is to be computed.
matrix	a logical indicating whether the whole matrix of values for the measure for each of the input track's subtracks is to be returned. Otherwise only the mean is returned.
min.segments	the number of segments that each regarded subtrack should at least consist of. Typically, this value would be set to the minimum number of segments that a (sub)track must have in order for the measure to be decently computed. For example, at least two segments are needed to compute the overallAngle.

#### Details

The measure is computed for each of the input track's subtracks of length at least min.segments, and the resulting values are either returned in a matrix (if matrix is set), or their mean is returned. The computed matrix is symmetric since the direction along which a track is traversed is assumed not to matter. The values at [i, i + j], where j is a nonnegative integer with  $j < \min$ .segments, (with the default value min.segments=1 this is exactly the main diagonal) are set to NA.

## Value

If matrix is set, a matrix with the values of the measure for all the input track's subtracks is returned. The value of this matrix at position [i, j] corresponds to the subtrack that starts with the input track's *j*th point and ends at its *i*th. Thus, with increasing column number, the regarded subtrack's starting point is advanced on the original track, and the values for increasingly long subtracks starting from this point can be found columnwise below the main diagonal, respectively. If 'matrix' is not set, the mean over the values of the measure for all subtracks of at least 'min.segments' segments is retruned.

## Examples

```
## Compute the staggered matrix for overallAngle applied to all long enough
## subtracks of the first T cell track
applyStaggered(TCells[[1]], overallAngle, matrix=TRUE, min.segments = 2)
```

as.data.frame.tracks Convert Tracks to Data Frame

## Description

Converts tracks from the list-of-matrices format, which is good for efficient processing and therefore the default in this package, to a single dataframe which is convenient for plotting or saving the data.

#### Usage

```
## S3 method for class 'tracks'
as.data.frame(
    x,
    row.names = NULL,
    optional = FALSE,
    include.timepoint.column = FALSE,
    idsAsFactors = TRUE,
    ...
)
```

16

## as.list.tracks

#### Arguments

х	the tracks object to be coerced to a data frame.	
row.names	NULL or a character vector giving row names for the data frame. Missing values are not allowed.	
optional	logical. Required for S3 consistency, but has no effect: column names are always assigned to the resulting data frame regardless of the setting of this option.	
include.timepoint.column		
	logical. If set to TRUE, then the resulting dataframe will contain a column that consecutively numbers the positions according to their time. Note that this information is anyway implicitly present in the time information.	
idsAsFactors	logical. If TRUE, then the id column of the resulting dataframe will be a factor column, otherwise a characeter column.	
	further arguments to be passed from or to other methods.	

## Value

A single data frame containing all individual tracks from the input with a prepended column named "id" containing each track's identifier in 'x'.

#### Examples

## Display overall average position of the T cell data colMeans( as.data.frame( TCells )[-c(1,2)] )

as.list.tracks	Convert from Tracks to List
	•/

## Description

Coerces a tracks object to a list.

## Usage

## S3 method for class 'tracks'
as.list(x, ...)

## Arguments

x	the tracks object to be coerced to a list.
	further arguments to be passed from or to other methods.

## Value

A generic list of single tracks, where each track is a matrix with t/delta.t rows and 4 columns. This looks a lot like a tracks object, except that its class is not "tracks" anymore.

as.tracks.data.frame Convert from Data Frame to Tracks

## Description

Get cell tracks from a data.frame. Data are expected to be organized as follows. One column contains a track identifier, which can be numeric or a string, and determines which points belong to the same track. Another column is expected to contain a time index or a time period (e.g. number of seconds elapsed since the beginning of the track, or since the beginning of the experiment). Input of dates is not (yet) supported, as absolute time information is frequently not available. Further columns contain the spatial coordinates. If there are three or less spatial coordinates, their names will by "x", "y", and "z" (depending on whether the tracks are 1D, 2D or 3D). If there are four or more spatial coordinates, their names will be "x1", "x2", and so on. The names or indices of these columns in the data.frame are given using the corresponding parameters (see below). Names and indices can be mixed, e.g. you can specify id.column="Parent" and pos.columns=1:3"

## Usage

```
## S3 method for class 'data.frame'
as.tracks(
    x,
    id.column = 1,
    time.column = 2,
    pos.columns = c(3, 4, 5),
    scale.t = 1,
    scale.pos = 1,
    ...
)
```

## Arguments

х	the data frame to be coerced to a tracks object.
id.column	index or name of the column that contains the track ID.
time.column	index or name of the column that contains elapsed time.
pos.columns	vector containing indices or names of the columns that contain the spatial coordinates. If this vector has two entries and the second entry is NA, e.g. $c('x', NA)$ or $c(5, NA)$ then all columns from the indicated column to the last column are used. This is useful when reading files where the exact number of spatial dimensions is not known beforehand.
scale.t	a value by which to multiply each time point. Useful for changing units, or for specifying the time between positions if this is not contained in the file itself.
scale.pos	a value, or a vector of values, by which to multiply each spatial position. Useful for changing units.
	further arguments to be passed to read.csv, for instance sep="\t" can be use- ful for tab-separated files.

#### **BCells**

#### Value

A tracks object.

BCells

Two-Photon Data: B Cells in a Lymph Node

#### Description

Labelled B cells were adoptively transfered and intravitally imaged (using two-photon microscopy) inside a peripheral lymph node of the recipient mouse. These data illustrate the characteristic "random-walk-like" motion pattern of B cells in lymph nodes.

#### Usage

data("BCells")

#### Format

An S3 object of class "tracks"; a list with 24 elements. Each element name identifies a cell track. Each element is a matrix containing the following four columns.

- t the time (in seconds)
- x The X coordinate (in micrometers)
- y The Y coordinate (in micrometers)
- z The Z coordinate (in micrometers)

#### Source

Data were generated in 2012 in the Mark J. Miller Lab, Department of Medicine, Washington University in St Louis, USA.

## References

Zinselmeyer BH, Dempster J, Wokosin DL, Cannon JJ, Pless R, Parker I and Miller MJ (2009), Two-photon microscopy and multi-dimensional analysis of cell dynamics. *Methods in Enzymology*, **461**:349–78. doi:10.1016/S0076-6879(09)05416-0

Konjufca V and Miller MJ (2009), Imaging Listeria monocytogenes infection in vivo. *Current Topics in Microbiology and Immunology*, **334**:199–226. doi:10.1007/978-3-540-93864-4\_9

Kreisel D, Nava RG, Li W, Zinselmeyer BH, Wang B, Lai J, Pless R, Gelman AE, Krupnick AS, and Miller MJ (2010), In vivo two-photon imaging reveals monocyte-dependent neutrophil extravasation during pulmonary inflammation. *PNAS*, **107**(42):18073–18078. doi:10.1073/pnas.1008737107

#### Examples

```
## load the tracks
data(BCells)
## visualize the tracks (calls function plot.tracks)
plot(BCells)
```

beaucheminTrack

#### Description

The Beauchemin model is a simple, particle-based description of T cell motion in lymph node in the absence of antigen, which is similar to a random walk (Beauchemin et al, 2007).

## Usage

```
beaucheminTrack(
   sim.time = 10,
   delta.t = 1,
   p.persist = 0,
   p.bias = 0.9,
   bias.dir = c(0, 0, 0),
   taxis.mode = 1,
   t.free = 2,
   v.free = 18.8,
   t.pause = 0.5
)
```

#### Arguments

sim.time	specifies the duration of the track to be generated
delta.t	change in time between each timepoint.
p.persist	indicates how probable a change in direction is. With p.persist = 1, the direction never changes between steps and with p.persist = 0, a new direction is sampled at every step.
p.bias	strength of movement in the direction of bias.dir.
bias.dir	a 3D vector indicating the direction along which there is a preference for movement.
taxis.mode	specified mode of movement. 1 := orthotaxis, 2 := topotaxis, 3 := klinotaxis.
t.free	time interval for how long the cell is allowed to move between turns.
v.free	speed of the cell during the free motion.
t.pause	time that it takes the cell to adjust movement to new direction.

#### Details

In the Beauchemin model, cells move into a fixed direction for a fixed time t.free at a fixed speed v.free. They then switch to a different direction, which is sampled at uniform from a sphere. The change of direction takes a fixed time t.pause, during which the cell does not move. Thus, the Beauchemin model is identical to the freely jointed chain model of polymer physics, except for the explicit "pause phase" between subsequent steps.

#### bootstrapTrack

The default parameters implemented in this function were found to most accurately describe 'default' T cell motion in lymph nodes using least-squares fitting to the mean displacement plot (Beauchemin et al, 2007).

This function implements an extended version of the Beauchemin model, which can also simulate directionally biased motion. For details, see Textor et al (2013).

## Value

A track, i.e., a matrix with t/delta.t rows and 4 columns.

## References

Catherine Beauchemin, Narendra M. Dixit and Alan S. Perelson (2007), Characterizing T cell movement within lymph nodes in the absence of antigen. *Journal of Immunology* **178**(9), 5505-5512. doi:10.4049/jimmunol.178.9.5505

Johannes Textor, Mathieu Sinn and Rob J. de Boer (2013), Analytical results on the Beauchemin model of lymphocyte migration. *BMC Bioinformatics* **14**(Suppl 6), S10. doi:10.1186/1471-2105-14-S6-S10

## Examples

bootstrapTrack	Simulate Tracks via Bootstrapping of Speed and Turning Angle from a
	Real Track Dataset

## Description

Returns a simulated dataset by sampling from the speed and turning angle distributions from an original track dataset (only in 2 or 3 dimensions)

#### Usage

bootstrapTrack(nsteps, trackdata)

boundingBox

#### Arguments

nsteps	desired number of steps (e.g. 10 steps generates a track with 11 positions).
trackdata	a tracks object to extract speeds and turning angles from.

## Details

The number of dimensions is kept the same as in the original data (if data is 3D but simulated tracks should be 2D, consider calling projectDimensions on the input data before supplying it to bootstrapTrack). The time interval between "measurements" of the simulated track equals that in the real data and is found via timeStep. The first step starts at the origin in a random direction, with a speed sampled from the speed distribution to determine its displacement. All subsequent steps also have their turning angles sampled from the turning angle distribution in the data.

#### Value

A data frame containing in cell track with nsteps steps in the same number of dimensions as the original data is returned.

## Generate bootstrapped tracks of the TCell data; compare its speed distribution to the ## original data (should be the same). T.bootstrap <- bootstrapTrack( 100, TCells ) step.speeds.real <- sapply( subtracks(TCells,1), speed ) step.speeds.bootstrap <- sapply( subtracks( T.bootstrap, 1), speed ) qqplot( step.speeds.real, step.speeds.bootstrap )

boundingBox Bounding Box of a Tracks Object

#### Description

Computes the minimum and maximum coordinates per dimension (including time) for all positions in a given list of tracks.

#### Usage

boundingBox(x)

#### Arguments

x the input tracks object.

#### Value

Returns a matrix with two rows and d + 1 columns, where d is the number of spatial dimensions of the tracks. The first row contains the minimum and the second row the maximum value of any track in the dimension given by the column.

## brownianTrack

#### Examples

```
## Use bounding box to set up plot window
bb <- boundingBox(c(TCells,BCells,Neutrophils))
plot( Neutrophils, xlim=bb[,"x"], ylim=bb[,"y"], col=1 )
plot( BCells, col=2, add=TRUE )
plot( TCells, col=3, add=TRUE )
```

brownianTrack

#### Simulate an Uncorrelated Random Walk

## Description

Generates a random track with nsteps steps in dim dimensions.

## Usage

brownianTrack(nsteps = 100, dim = 3, mean = 0, sd = 1)

## Arguments

nsteps	desired number of steps (e.g. 10 steps generates a track with 11 positions).
dim	desired number of dimensions.
mean	stepwise mean drift per dimension; use 0 for an unbiased Brownian motion and other values for Brownian motion with drift.
sd	stepwise standard deviation per dimension.

## Details

In in every step an for each dimension, a normally distributed value with mean mean and standard deviation sd is added to the previous cell position.

#### Value

A data frame containing in cell track with nsteps steps in dim dimensions is returned.

## The Hurst exponent of a 1D Brownian track should be near 0.5 hurstExponent( brownianTrack( 100, 1))

cellPairs

#### Description

Get all unique combinations of two track ids.

#### Usage

cellPairs(X)

#### Arguments

X a tracks object

#### Value

A dataframe with two columns: one for each of the track ids in the pair. Each row represents a pair.

#### Examples

```
## Find all pairs of cells in the T cell data
pairs <- cellPairs( TCells )</pre>
```

celltrackR

celltrackR: Quantitative analysis of motion.

#### Description

The CelltrackR package is designed for analyzing cell tracks acquired by time-lapse microscopy (like those provided in the included datasets TCells, BCells and Neutrophils). But it can of course process any x-y-(z)-t data, and we hope that it may be useful for other purposes as well.

#### Details

For a complete list of functions, use help(package="celltrackR"). A handy cheat sheet is available in pdf. You can open it by calling the function cheatsheet.

## Data structure

The basic data structure that most functions in this package operate on is a set of *tracks*. A track is a list of spatial coordinates that are recorded at *fixed* time intervals; the function timeStep can be used to check for fluctuations of the recording intervals.

We expect tracks to be stored in a matrix (or data frame, but this is discouraged for efficiency reasons) whose first column denotes a time interval (e.g. seconds elapsed since the beginning of the experiment), and whose remaining columns denote a spatial coordinate. A set of tracks is stored as a list with S3 class tracks. CelltrackR provides some S3 methods for this class, which are explained in tracks as well as plot.tracks, sort.tracks and as.list.tracks.

#### celltrackR

#### Track analysis in celltrackR

A wide range of common track measures are included in the package. These are all functions that take a single track as an input, and output one or several numbers. For instance, the function speed estimates the average instantaneous speed of the track by linear interpolation, and straightness computes the start-to-end distance divided by the trajectory length (a number between 0 and 1, where 1 indicates a perfectly straight track). See TrackMeasures for an overview of measures that can be analyzed on tracks. Also see AngleAnalysis for an overview of functions that can help detect directional bias and tracking artefacts (see Beltman et al, 2009).

CelltrackR is designed to support various flavors of track analysis that have been suggested in the literature. The simplest kind is a *track-based* analysis, where we compute a single statistic for each track in a dataset (Beltman et al, 2009). Because track sets are lists, this is achieved simply by using lapply or sapply together with the track measure (see Examples).

In *step-based* analyses (Beltman et al, 2009), we chop each track up into segments of the same length and then apply our measures to those segments. This can help to avoid biases that arise from variations in track length (which are always present in cell tracking experiments). In CelltrackR, step-based analyses are performed by using the subtracks function. Often we want to perform such step-based analyses for all possible subtrack lengths simultaneously, and plot the result as a function of the subtrack length; a famous example is the *mean square displacement plot*. This can be achieved by using the aggregate.tracks function, which has options to control which subtrack lengths are considered and whether overlapping subtracks are considered.

In a *staggered staggered* analysis (Mokhtari et al, 2013), we analyse all subtracks (of any length) of a single track, and typically plot the result as a matrix. This can reveal dynamic patterns along a single track, e.g. turning behaviour or local slowdowns. Staggered analyses can be performed using the applyStaggered function.

#### Simulating tracks in celltrackR

Lastly, in addition to data analysis, the package contains some function to generate cell tracks by simulation. This is useful to develop and benchmark track analysis methodology (Textor et al, 2011), and for computational biology studies that try to extrapolate the long-term consequences of observed cell migration behaviour. Alongside a simple uncorrelated random walk (brownianTrack), this package implements a simulation model proposed by Beauchemin et al (2007) in the function beaucheminTrack. That model can also simulate directionally biased motion.

## Author(s)

Johannes Textor, Katharina Dannenberg, Jeffrey Berry, Gerhard Burger, Inge Wortel Maintainer: Johannes Textor <johannes.textor@gmx.de>

#### References

Joost B. Beltman, Athanasius F.M. Maree and Rob. J. de Boer (2009), Analysing immune cell migration. *Nature Reviews Immunology* **9**, 789–798. doi:10.1038/nri2638

Zeinab Mokhtari, Franziska Mech, Carolin Zitzmann, Mike Hasenberg, Matthias Gunzer and Marc Thilo Figge (2013), Automated Characterization and Parameter–Free Classification of Cell Tracks Based on Local Migration Behavior. *PLoS ONE* **8**(12), e80808. doi:10.1371/journal.pone.0080808

Johannes Textor, Antonio Peixoto, Sarah E. Henrickson, Mathieu Sinn, Ulrich H. von Andrian and Juergen Westermann (2011), Defining the Quantitative Limits of Intravital Two-Photon Lymphocyte Tracking. *PNAS* **108**(30):12401–12406. doi:10.1073/pnas.1102288108

Catherine Beauchemin, Narendra M. Dixit and Alan S. Perelson (2007), Characterizing T cell movement within lymph nodes in the absence of antigen. *Journal of Immunology* **178**(9), 5505-5512. doi:10.4049/jimmunol.178.9.5505

## See Also

The package vignettes, available from browseVignettes(package="celltrackR"). Make sure you have installed the package with option build\_vignettes = TRUE, or vignettes will not be visible. Also check out the package cheat sheet, which is available by calling the function cheatsheet.

#### Examples

```
## track-based speed comparison
boxplot(sapply( Neutrophils, straightness ), sapply( BCells, straightness ))
## step-based turning angle comparison
boxplot(sapply(subtracks(Neutrophils, 2), overallAngle),
   sapply(subtracks(BCells, 2), overallAngle))
## mean square displacement plot; a step-based displacement analysis for all step lengths
plot(aggregate(TCells, squareDisplacement)[,"value"])
## 'staggered' analysis of displacement over whole track. Reveals that this track
## slows down near its beginning and near its end.
filled.contour(applyStaggered(TCells[[4]], displacement, matrix=TRUE))
## a simple hierarchical clustering based on 2D asphericity
## tag track IDs so we can identify them later
names(TCells) <- paste0("T",names(TCells))</pre>
names(BCells) <- paste0("B",names(BCells))</pre>
names(Neutrophils) <- paste0("N",names(Neutrophils))</pre>
## project all tracks down to 2D
cells <- projectDimensions(c(TCells,BCells,Neutrophils), c("x","y"))</pre>
## compute asphericity
asph <- lapply(cells, asphericity)</pre>
## plot clustering
plot(hclust(dist(asph)))
```

cheatsheet

## clusterTracks

## Description

Running this function will open the package cheat sheet (a pdf) via a call to system().

## Usage

```
cheatsheet(opencmd = NULL)
```

## Arguments

opencmd The command used to open pdfs from the command line.

## Value

None

clusterTracks Cluster Tracks

## Description

Perform a quick clustering visualization of a set of tracks according to a given vector of track measures.

## Usage

```
clusterTracks(
  tracks,
  measures,
  scale = TRUE,
  labels = NULL,
  method = "hclust",
  return.clust = FALSE,
  ...
)
```

#### Arguments

tracks	the tracks that are to be clustered.
measures	a function, or a vector of functions (see TrackMeasures). Each function is expected to return a single number given a single track.
scale	logical indicating whether the measures values shall be scaled using the function scale before the clustering.
labels	optional: a vector of labels of the same length as the track object. These are used to color points in the visualization.
method	"hclust" for hierarchical clustering, or "kmeans" for k-means clustering.
return.clust	logical: return the clustering object instead of only the plot? (defaults to FALSE).
	additional parameters to be passed to the corresponding clustering function: hclust or kmeans.

#### Details

The measures are applied to each of the tracks in the given *tracks* object. According to the resulting values, the tracks are clustered using the chosen clustering method. If scale is TRUE, the measure values are scaled to mean value 0 and standard deviation 1 (per measure) before the clustering.

Method hclust plots a dendrogram of the clustering.

Method kmeans plots each computed cluster (x-axis) versus each of the track measures in the measures vector, producing one panel per measure. If labels are given, points are colored according to their "true" label.

## Value

By default, only returns a plot. If return.clust=TRUE, also returns a clustering object as returned by hclust or kmeans. output object.

#### See Also

getFeatureMatrix to obtain a feature matrix that can be used for manual clustering and plotting, and trackFeatureMap to visualize high-dimensional track feature data via dimensionality reduction.

#### Examples

## Cluster tracks according to the mean of their Hust exponents along X and Y
## using hierarchical clustering

```
cells <- c(TCells,Neutrophils)
real.celltype <- rep(c("T","N"),c(length(TCells),length(Neutrophils)))
## Prefix each track ID with its cell class to evaluate the clustering visually
names(cells) <- paste0(real.celltype,seq_along(cells))
clust <- clusterTracks( cells, hurstExponent, method = "hclust",
    return.clust = TRUE )
</pre>
```

## How many cells are "correctly" clustered?
sum( real.celltype == c("T","N")[cutree(clust,2)] )

distanceCells Minimum Distance between Two Cells

## Description

Compute the minimum distance between two cells in the dataset (minimum over all) the timepoints where they were both measured.

#### Usage

distanceCells(X, cellids)

#### distanceSteps

#### Arguments

Х	a tracks object
cellids	a vector of two indices specifying the tracks to compute distance between.

#### Value

A single distance, or NA if the the tracks do not have overlapping timepoints.

#### See Also

angleCells to compute the angle between the track displacement vectors, and AngleAnalysis for other methods to compute angles and distances.

## Examples

```
## Find the minimum distance between the tracks with ids 1 and 2 distanceCells( TCells, c("1","2") )
```

distanceSteps

Distance between Two Steps

#### Description

Compute the distance between two steps in the dataset that occur at the same timepoint. The distance is the distance between the step starting points.

#### Usage

```
distanceSteps(X, trackids, t, quietly = FALSE)
```

## Arguments

Х	a tracks object
trackids	a vector of two indices specifying the tracks to get steps from.
t	the timepoint at which the steps should start.
quietly	logical; should a warning be returned if one or both of the steps are missing in the data and the function returns NA?

## Value

A single distance, or NA if the desired timepoint is missing for one or both of the tracks.

#### See Also

angleSteps to compute the angle between the steps, timePoints to list all timepoints in a dataset, and AngleAnalysis for other methods to compute angles and distances.

## Examples

```
## Find the distance between the steps of the tracks with ids 1 and 2, at the 3rd
## timepoint in the dataset.
t <- timePoints( TCells )[3]
distanceSteps( TCells, c("1","2"), t )</pre>
```

distanceToPlane Distance to a Reference Plane

#### Description

Compute the (shortest) distance between the starting point of a track and a reference plane. Useful to detect directed movement and/or tracking artefacts.

#### Usage

```
distanceToPlane(x, p1 = c(0, 0, 0), p2 = c(0, 1, 0), p3 = c(1, 0, 0), from = 1)
```

#### Arguments

x	a single input track; a matrix whose first column is time and whose remaining columns are a spatial coordinate.
p1, p2, p3	numeric vectors of coordinates of three points specifying a reference plane to compute distances to.
from	index, or vector of indices, of the first row of the track. If from is a vector, distances are returned for all steps starting at the indices in from.

## Value

A single distance.

#### See Also

angleToPlane to compute the angle to the plane, and AngleAnalysis for other methods to compute angles and distances.

#### Examples

## Plotting the angle versus the distance to a reference plane can be informative to ## detect tracking artefacts near the border of the imaging volume. ## We should be suspicious especially when small angles are more frequent at low distances ## to the border planes. steps <- subtracks( TCells, 1 ) minz <- boundingBox( TCells )["min","z"] ## Compute angles and distances to the lower plane in z-dimension angles <- sapply( steps, angleToPlane, p1 = c(0,0,minz), p2 = c(1,0,minz), p3 = c(0,1,minz) ) distances <- sapply( steps, distanceToPlane, p1 = c(0,0,minz), p2 = c(1,0,minz), p3 = c(0,1,minz) ) scatter.smooth( distances, angles ) abline( h = 32.7, col = "red" )

30

#### Description

Compute the distance between the starting point of a track and a reference point. Useful to detect directed movement towards a point (see examples).

#### Usage

distanceToPoint(x, p = c(0, 0, 0), from = 1)

#### Arguments

x	a single input track; a matrix whose first column is time and whose remaining columns are a spatial coordinate.
р	numeric vector of coordinates of the reference point p to compute distances to.
from	index, or vector of indices, of the first row of the track. If from is a vector, distances are returned for all steps starting at the indices in from.

## Value

A single distance.

## See Also

angleToPoint to compute the angle to the reference point, and AngleAnalysis for other methods to compute angles and distances.

## Examples

```
## Plotting the angle versus the distance to a reference point can be informative to
## detect biased movement towards that point. We should be suspicious especially
## when small angles are more frequent at lower distances.
steps <- subtracks( Neutrophils, 1 )
bb <- boundingBox( Neutrophils )
angles <- sapply( steps, angleToPoint, p = bb["max",-1] )
distances <- sapply( steps, distanceToPoint, p = bb["max",-1] )
scatter.smooth( distances, angles )
abline( h = 90, col = "red" )</pre>
```

filterTracks

## Description

Extracts subtracks based on a given function.

## Usage

filterTracks(f, x, ...)

## Arguments

f	a function that accepts a single track as its first argument and returns a logical
	value (or a value that can be coerced to a locical).
х	a tracks object.
	further arguments to be passed on to f.

## Value

A tracks object containing only those tracks from x for which f evaluates to TRUE.

#### Examples

```
## Remove short tracks from the T cells data
plot( filterTracks( function(t) nrow(t)>10, TCells ) )
```

getFeatureMatrix Obtaining A Feature Matrix

## Description

Applies a given vector of track measures directly on a set of tracks, returning output in a matrix with a column for each measure and a row for each track. Can also return a distance matrix, which some clustering methods require.

## Usage

```
getFeatureMatrix(tracks, measures, dist = FALSE, ...)
```

#### Arguments

tracks	the tracks that are to be analyzed.
measures	a function, or a vector of functions (see TrackMeasures). Each function is expected to return a single number given a single track.
dist	should a distance matrix rather than a feature matrix be returned?
	further arguments passed on to "dist"

## hotellingsTest

## Value

A matrix with a row for each track and a column for each measure.

## See Also

clusterTracks for a quick method to compute the feature matrix and a clustering, and trackFeatureMap to perform dimensionality reduction methods on a set of track features.

## Examples

```
## Get speed, meanTurningAngle, and straightness for T cell tracks
fm <- getFeatureMatrix( TCells, c(speed,meanTurningAngle,straightness))
str(fm)
```

hotellingsTest Test Unbiasedness of Motion

#### Description

Test the null hypothesis that a given set of tracks originates from an uncorrelated and unbiased type of motion (e.g., a random walk without drift). This is done by testing whether the mean step vector is equal to the null vector.

#### Usage

```
hotellingsTest(
  tracks,
  dim = c("x", "y"),
  step.spacing = 0,
  plot = FALSE,
  add = FALSE,
  ellipse.col = "blue",
  ellipse.border = "black",
  conf.level = 0.95,
  ...
)
```

## Arguments

tracks	the tracks whose biasedness is to be determined.
dim	vector with the names of the track's dimensions that are to be considered. By default $c("x", "y")$ .
step.spacing	How many positions are to be left out between the steps that are considered for the test. For persistent motion, subsequent steps will be correlated, which leads to too low p-values because Hotelling's test assumes that the input data is independent. To avoid this, the resulting p-value should either be corrected

	for this dependence (e.g. by adjusting the degrees of freedom accordingly), or 'step.spacing' should be set to a value high enough to ensure that the considered steps are approximately independent.
plot	logical indicating whether the scatter of the step's directions, origin of ordinates (green circle) and the mean of the data points (green cross) are to be plotted. (In one dimension also the bounds of the condfidence interval are given.) Plot works only in one or two dimensions.
add	whether to add the plot to the current plot (TRUE) or create a
ellipse.col	color with which to draw the confidence ellipse of the mean (for 1D, this corresponds to the confidence interval of the mean). Use NA to omit the confidence ellipse.
ellipse.border	color of the confidence ellipse border. Use NA to omit the border.
conf.level	the desired confidence level for the confidence ellipse.
	further arguments passed on to plot.

#### Details

Computes the displacement vectors of all segments in the tracks given in tracks, and performs Hotelling's T-square Test on that vector.

#### Value

A list with class htest.

#### References

Johannes Textor, Antonio Peixoto, Sarah E. Henrickson, Mathieu Sinn, Ulrich H. von Andrian and Juergen Westermann (2011), Defining the Quantitative Limits of Intravital Two-Photon Lymphocyte Tracking. PNAS 108(30):12401-12406. doi:10.1073/pnas.1102288108

#### Examples

## Test  $H_0$ : T-cells migrate by uncorrelated random walk on x and y coordinates, ## and report the p-value. hotellingsTest( TCells )\$p.value

interpolateTrack Interpolate Track Positions

#### Description

Approximates the track positions at given time points using linear interpolation (via the approx function).

## maxTrackLength

## Usage

interpolateTrack(x, t, how = "linear")

## Arguments

x	the input track (a matrix or data frame).
t	the times at which to approximate track positions. These must lie within the interval spanned by the track timepoints.
how	specifies how to perform the interpolation. Possible values are "linear" (which uses approx with default values) and "spline" (which uses spline with default values).

## Value

The interpolated track (a matrix or data frame).

## Examples

```
## Compare interpolated and non-interpolated versions of a track
bb <- boundingBox( TCells[2] )
plot( TCells[2] )
t2i <- interpolateTrack(TCells[[2]], seq(bb[1,"t"],bb[2,"t"],length.out=100),"spline")
plot( tracks( t2i ), add=TRUE, col=2 )
```

maxTrackLength Length of Longest Track

## Description

Determines the maximum number of positions over the tracks in x.

## Usage

```
maxTrackLength(x)
```

## Arguments

х

the tracks object the tracks in which are to be considered.

## Value

The maximum number of rows of a track in x

```
Neutrophils
```

#### Description

Labelled neutrophils were adoptively transfered and intravitally imaged (using two-photon microscopy) inside the lung of the recipient mouse. These cells display a fairly directed kind of motion, as they move towards infection foci.

#### Usage

```
data("Neutrophils")
```

## Format

An S3 object of class "tracks"; a list with 10 elements. Each element name identifies a cell track. Each element is a matrix containing the following four columns.

- t the time (in seconds)
- x The X coordinate (in micrometers)
- y The Y coordinate (in micrometers)
- z The Z coordinate (in micrometers)

#### Source

Data were generated in 2012 in the Mark J. Miller Lab, Department of Medicine, Washington University in St Louis, USA.

## References

Zinselmeyer BH, Dempster J, Wokosin DL, Cannon JJ, Pless R, Parker I and Miller MJ (2009), Two-photon microscopy and multi-dimensional analysis of cell dynamics. *Methods in Enzymology*, **461**:349–78. doi:10.1016/S0076-6879(09)05416-0

Konjufca V and Miller MJ (2009), Imaging Listeria monocytogenes infection in vivo. *Current Topics in Microbiology and Immunology*, **334**:199–226. doi:10.1007/978-3-540-93864-4\_9

Kreisel D, Nava RG, Li W, Zinselmeyer BH, Wang B, Lai J, Pless R, Gelman AE, Krupnick AS, and Miller MJ (2010), In vivo two-photon imaging reveals monocyte-dependent neutrophil extravasation during pulmonary inflammation. *PNAS*, **107**(42):18073–18078. doi:10.1073/pnas.1008737107

#### Examples

```
## load the tracks
data(Neutrophils)
## visualize the tracks (calls function plot.tracks)
plot(Neutrophils)
```

normalizeToDuration Normalize a Measure to Track Duration

#### Description

Returns a measure that divides the input measure by the duration of its input track.

## Usage

```
normalizeToDuration(x)
```

#### Arguments

х

a track measure (see TrackMeasures).

## Value

A function that computes the input measure for a given track and returns the result divided by the track's duration.

#### Examples

```
## normalizeToDuration(displacement) can be used as an indicator
## for the motion's efficiency
sapply(TCells, normalizeToDuration(displacement))
```

normalizeTracks Normalize Tracks

## Description

Translates each track in a given set of tracks such that the first position is the origin.

## Usage

```
normalizeTracks(x)
```

## Arguments

х

the input tracks object.

#### Value

an output tracks object with all tracks shifted such that their starting position lies at the origin of the coordinate system.

## Examples

```
## normalization of Neutrophil data reveals upward motion
plot( normalizeTracks( Neutrophils ) )
```

plot.tracks Plot Tracks in 2D

## Description

Plots tracks contained in a "tracks" object into a twodimensional space pallelel to the data's axes.

#### Usage

```
## S3 method for class 'tracks'
plot(
    x,
    dims = c("x", "y"),
    add = F,
    col = order(names(x)),
    pch.start = 1,
    pch.end = NULL,
    cex = 0.5,
    ...
)
```

#### Arguments

х	the tracks to be plotted.
dims	a vector giving the dimensions of the track data that shall be plotted, e.g. $c('x', 'y')$ for the x and y dimension.
add	boolean value indicating whether the tracks are to be added to the current plot.
col	a specification of the $color(s)$ to be used. This can be a vector of size $length(x)$ , where each entry specififes the color for the corresponding track.
pch.start	point symbol with which to label the first position of the track (see points).
pch.end	point symbol with which to label the last position of the track
cex	point size for positions on the tracks.
	additional parameters (e.g. xlab, ylab). to be passed to plot (for add=FALSE) or points (for add=TRUE), respectively.

## Details

One dimension of the data (by default y) is plotted against another (by default x). The dimesions can be chosen by means of the parameter dims and the axes can be labeled accordingly with the aid of xlab and ylab. The color can be set through col. If the tracks should be added to an existing plot, add is to be set to TRUE.

38

## plot3d

## Value

None

## See Also

plot3d

plot3d

Plot Tracks in 3D

## Description

Takes an input tracks object and plots them in 3D using the scatterplot3d function.

## Usage

plot3d(x, ...)

## Arguments

х	the tracks which will be plotted in 3d
	further arguments to be passed on to scatterplot3d

## Value

None.

## Examples

```
if( require("scatterplot3d",quietly=TRUE) ){
   plot3d( TCells )
}
```

plotTrackMeasures Bivariate Scatterplot of Track Measures

## Description

Plots the values of two measures applied on the given tracks against each other.

## Usage

```
plotTrackMeasures(
    x,
    measure.x,
    measure.y,
    add = FALSE,
    xlab = deparse(substitute(measure.x)),
    ylab = deparse(substitute(measure.y)),
    ellipse.col = "red",
    ellipse.border = "black",
    conf.level = 0.95,
    ...
)
```

## Arguments

х	the input tracks object.
measure.x	the measure to be shown on the X axis (see TrackMeasures).
measure.y	the measure to be shown on the Y axis.
add	a logical indicating whether the tracks are to be added to an existing plot via points.
xlab	label of the x-axis. By default the name of the input function measure.x.
ylab	label of the y-axis. By default the name of the input function measure.y.
ellipse.col	color with which to draw the confidence ellipse of the mean (for 1D, this corresponds to the confidence interval of the mean). Use NA to omit the confidence ellipse.
ellipse.border	color of the confidence ellipse border. Use NA to omit the border.
conf.level	the desired confidence level for the confidence ellipse.
	additional parameters to be passed to plot (in case add=FALSE) or points (add=TRUE).

#### Details

Plots the value of measurey applied to x against the value of measurey applied to y. This is useful for "FACS-like" motility analysis, where clusters of cell tracks are identified based on their motility parameters (Moreau et al, 2012; Textor et al, 2014).

## Value

None

## References

Moreau HD, Lemaitre F, Terriac E, Azar G, Piel M, Lennon-Dumenil AM, Bousso P (2012), Dynamic In Situ Cytometry Uncovers T Cell Receptor Signaling during Immunological Synapses and Kinapses In Vivo. *Immunity* **37**(2), 351–363. doi:10.1016/j.immuni.2012.05.014

40

#### prefixes

Johannes Textor, Sarah E. Henrickson, Judith N. Mandl, Ulrich H. von Andrian, J\"urgen Westermann, Rob J. de Boer and Joost B. Beltman (2014), Random Migration and Signal Integration Promote Rapid and Robust T Cell Recruitment. *PLoS Computational Biology* **10**(8), e1003752. doi:10.1371/journal.pcbi.1003752

#### Examples

```
## Compare speed and straightness of 3 example population tracks.
## To make the comparison fair, analyze subtracks of fixed length.
plotTrackMeasures( subtracks(TCells,4,0), speed, straightness, ellipse.col="black" )
plotTrackMeasures( subtracks(BCells,4,0), speed, straightness,
    col=2, ellipse.col=2, pch=2, add=TRUE )
plotTrackMeasures( subtracks(Neutrophils,4,0), speed, straightness,
    col=3, ellipse.col=3, pch=3, add=TRUE )
```

prefixes

Get Track Prefixes

#### Description

Creates a tracks object consisting of all prefixes (i.e., subtracks starting with the first position of a track) of 'x' with 'i' segments (i.e., 'i'+1 positions).

#### Usage

prefixes(x, i)

#### Arguments

Х	a single track or a tracks object.
i	subtrack length. A single integer, lists are not supported.

#### Details

This function behaves exactly like subtracks except that only subtracks starting from the first position are considered.

## Value

A *tracks* object is returned which contains all the subtracks of any track in the input *tracks* object that consist of exactly 'i' segments and start at the first registered coordinate of the given track.

#### See Also

subtracks to extract all subtracks of a given length, subtracksByTime to extract all subtracks of a given length starting at some fixed timepoint, and selectSteps to extract single steps starting at a fixed timepoint from a subset of trackids.

projectDimensions Extract Spatial Dimensions

#### Description

Projects tracks onto the given spatial dimensions.

#### Usage

```
projectDimensions(x, dims = c("x", "y"))
```

#### Arguments

х	the input tracks object.
dims	a character vector (for column names) or an integer vector (for column indices) giving the dimensions to extract from each track. The time dimension (i.e., the
	first column of all tracks) is always included.

## Value

A tracks object is returned that contains only those dimensions of the input tracks that are given in dims.

#### Examples

```
## Compare 2D and 3D speeds
speed.2D <- mean( sapply( subtracks( projectDimensions( TCells, c("x","z") ), 2 ), speed ) )
speed.3D <- mean( sapply( TCells, speed ) )</pre>
```

read.tracks.csv Read Tracks from Text File

## Description

Reads cell tracks from a CSV or other text file. Data are expected to be organized as follows. One column contains a track identifier, which can be numeric or a string, and determines which points belong to the same track. Another column is expected to contain a time index or a time period (e.g. number of seconds elapsed since the beginning of the track, or since the beginning of the experiment). Input of dates is not (yet) supported, as absolute time information is frequently not available. Further columns contain the spatial coordinates. If there are three or less spatial coordinates, their names will by "x", "y", and "z" (depending on whether the tracks are 1D, 2D or 3D). If there are four or more spatial coordinates, their names will be "x1", "x2", and so on. The names or indices of these columns in the CSV files are given using the corresponding parameters (see below). Names and indices can be mixed, e.g. you can specify id.column="Parent" and pos.columns=1:3

## read.tracks.csv

## Usage

```
read.tracks.csv(
   file,
   id.column = 1,
   time.column = 2,
   pos.columns = c(3, 4, 5),
   scale.t = 1,
   scale.pos = 1,
   header = TRUE,
   sep = "",
   track.sep.blankline = FALSE,
   ...
)
```

## Arguments

file	the name of the file which the data are to be read from, a readable text-mode connection or a complete URL (see read.table).	
id.column	index or name of the column that contains the track ID.	
time.column	index or name of the column that contains elapsed time.	
pos.columns	vector containing indices or names of the columns that contain the spatial coordinates. If this vector has two entries and the second entry is NA, e.g. $c('x', NA)$ or $c(5, NA)$ then all columns from the indicated column to the last column are used. This is useful when reading files where the exact number of spatial dimensions is not known beforehand.	
scale.t	a value by which to multiply each time point. Useful for changing units, or for specifying the time between positions if this is not contained in the file itself.	
scale.pos	a value, or a vector of values, by which to multiply each spatial position. Useful for changing units.	
header	a logical value indicating whether the file contains the names of the variables as its first line. See read.table.	
sep	a character specifying how the colums of the data are separated. The default value "" means columns are separated by tabs or other spaces. See read.table.	
track.sep.blankline		
	logical. If set to TRUE, then tracks are expected to be separated by one or more blank lines in the input file instead of being designated by a track ID column. In this case, numerical track IDs are automatically generated.	
	further arguments to be passed to read.csv, for instance $sep="\t"$ can be useful for tab-separated files.	

## Details

The input file's first four fields are interpreted as id, pos, t and x, respectively, and, if available, the fifth as y and the sixth as z. The returned object has the class *tracks*, which is a list of data frames representing the single tracks and having columns t and x, plus y and z, if necessary. The tracks' ids are retained in their position in the list, while the field pos will be unmaintained.

## Value

An object of class *tracks* is returned, which is a list of matrices, each containing the positions of one track. The matrices have a column t, followed by one column for each of the input track's coordinates.

repairGaps

Process Tracks Containing Gaps

## Description

Many common motility analyses, such as mean square displacement plots, assume that object positions are recorded at constant time intervals. For some application domains, such as intravital imaging, this may not always be the case. This function can be used to pre-process data imaged at nonconstant intervals, provided the deviations are not too extreme.

#### Usage

```
repairGaps(x, how = "split", tol = 0.05, split.min.length = 2)
```

#### Arguments

x	the input tracks object.
how	string specifying what do with tracks that contain gaps. Possible values are:
	• "drop": the simplest option – discard all tracks that contain gaps.
	• "split": split tracks around the gaps, e.g. a track for which the step between the 3rd and 4th positions is too long or too short is split into one track corresponding to positions 1 to 3 and another track corresponding to position 3 onwards.
	• "interpolate": approximate the track positions using linear interpolation (see interpolateTrack). The result is a tracks object with constant step durations.
tol	nonnegative number specifying by which fraction each step may deviate from the average step duration without being considered a gap. For instance, if the average step duration (see timeStep) is 100 seconds and tol is 0.05 (the de- fault), then step durations between 95 and 105 seconds (both inclusive) are not considered gaps. This option is ignored for how="interpolate".
split.min.length	
	nonnegative integer. For how="split", this discards all resulting tracks shorter than this many positions.

#### Value

A tracks object with gaps fixed according to the chosen method.

## selectSteps

#### Examples

```
## The Neutrophil data are imaged at rather nonconstant intervals
print( length( Neutrophils ) )
print( length( repairGaps( Neutrophils, tol=0.01 ) ) )
```

```
selectSteps
```

Get Single Steps Starting at a Specific Time from a Subset of Tracks

## Description

Obtain all single steps starting at a given timepoint t from a subset of tracks of interest.

#### Usage

```
selectSteps(X, trackids, t)
```

## Arguments

Х	Tracks object to obtain subtracks from
trackids	Character vector with the ids of tracks of interest
t	Timepoint at which the subtracks should start

## Value

A tracks object is returned which contains all the extracted steps.

#### See Also

subtracks to extract all subtracks of a given length, prefixes to extract all subtracks of a given length starting from the first coordinate in each track, subtracksByTime to extract all subtracks of a given length starting at some fixed timepoint, and timePoints to return all timepoints occurring in the dataset.

## Examples

```
## Get and plot all steps starting at the third timepoint in tracks 1 and 3 of
## the T cell dataset
subT <- selectSteps( TCells, c("1","5"), timePoints(TCells)[3] )
plot( subT )
```

selectTracks

#### Description

Given a tracks object, extract a subset based on upper and lower bounds of a certain measure. For instance, extract all tracks with a certain minimum length.

#### Usage

selectTracks(x, measure, lower, upper)

## Arguments

х	the input tracks.
measure	measure on which the selection is based (see TrackMeasures).
lower	specifies the lower bound (inclusive) of the allowable measure.
upper	specifies the upper bound (inclusive) of the allowable measure.

## Value

A tracks object with the selected subset of tracks.

## Examples

```
## Slower half of T cells
slow.tcells <- selectTracks( TCells, speed, -Inf, median( sapply(TCells, speed) ) )</pre>
```

simulateTracks Generate Tracks by Simulation

## Description

Generic function that executes expr, which is expected to return a track, n times and stores the output in a tracks object. Basically, this works like replicate but for tracks.

## Usage

simulateTracks(n, expr)

## Arguments

n	number of tracks to be generated.
expr	the expression, usually a call, that generates a single track.

## sort.tracks

## Value

A tracks object containing n tracks.

#### Examples

```
## Generate 10 tracks, 100 steps each, from a random walk with standard normally
## distributed increments and plot them
plot( simulateTracks( 10, brownianTrack(100,3) ) )
```

sort.tracks

## Sort Track Positions by Time

## Description

Sorts the positions in each track in a tracks object by time.

## Usage

## S3 method for class 'tracks'
sort(x, decreasing = FALSE, ...)

## Arguments

х	the <i>tracks</i> object whose tracks are to be sorted by time.
decreasing	logical. Should the sort be increasing or decreasing? Provided only for consis- tency with the generic sort method. The positions in each track should be sorted in increasing time order.
•••	further arguments to be passed on to order.

#### Details

Sorts the positions of each track (represented as a data frame) in the *tracks* object by time (given in the column t).

#### Value

A tracks object that contains the tracks from the input object sorted by time is returned.

splitTrack

#### Description

Split Track into Multiple Tracks

## Usage

splitTrack(x, positions, id = NULL, min.length = 2)

## Arguments

х	the input track (a data frame or a matrix).
positions	a vector of positive integers, given in ascending order.
id	a string used to identify the resulting tracks; for instance, if $id="X"$ , then the resulting tracks are named X_1, X_2 and so forth. Otherwise, they are simply labelled with integer numbers.
min.length	nonnegative integer. Resulting tracks that have fewer positions than the value of this parameter are dropped.

## Value

An object of class *tracks* with the resulting splitted tracks.

staggered	Staggered Version of a Function
-----------	---------------------------------

## Description

Returns the "staggered" version of a track measure. That is, instead of computing the measure on the whole track, the measure is averaged over all subtracks (of any length) of the track.

## Usage

```
staggered(measure, ...)
```

## Arguments

measure	a track measure (see TrackMeasures).
	further parameters passed on to applyStaggered

## Details

This is a wrapper mainly designed to provide a convenient interface for track-based staggered computations with lapply, see example.

## stepPairs

## Value

Returns a function that computes the given measure in a staggered fashion on that track.

## References

Zeinab Mokhtari, Franziska Mech, Carolin Zitzmann, Mike Hasenberg, Matthias Gunzer and Marc Thilo Figge (2013), Automated Characterization and Parameter–Free Classification of Cell Tracks Based on Local Migration Behavior. *PLoS ONE* **8**(12), e80808. doi:10.1371/journal.pone.0080808

#### Examples

```
hist( sapply( TCells, staggered( displacement ) ) )
```

stepPairs

Find Pairs of Steps Occurring at the Same Time

## Description

Find cell indices and timepoints where these cells both have a step.

#### Usage

```
stepPairs(X, filter.steps = NULL)
```

#### Arguments

Х	a tracks object
filter.steps	optional: a function used to filter steps on. See examples.

#### Value

A dataframe with three columns: two for the indices of cellpairs that share a step, and one for the timepoint at which they do so.

## Examples

```
## Find all pairs of steps in the T cell data that displace at least 2 microns. pairs <- stepPairs( TCells, filter.steps = function(t) displacement(t) > 2 )
```

subsample

## Description

Make tracks more coarse-grained by keeping only every kth position.

## Usage

subsample(x, k = 2)

## Arguments

х	an input track or tracks object.
k	a positive integer. Every $k$ th position of each input track is kept.

## Value

A tracks object with the new, more coarse-grained tracks.

#### See Also

interpolateTrack, which can be used for more flexible track coarse-graining.

#### Examples

```
## Compare original and subsampled versions of the T cell tracks
plot( TCells, col=1 )
plot( subsample( TCells, 3 ), col=2, add=TRUE, pch.start=NULL )
```

subtracks

Decompose Track(s) into Subtracks

## Description

Creates a tracks object consisting of all subtracks of 'x' with 'i' segments (i.e., 'i'+1 positions).

#### Usage

subtracks(x, i, overlap = i - 1)

#### Arguments

x	a single track or a tracks object.
i	subtrack length. A single integer, lists are not supported.
overlap	the number of segments in which each subtrack shall overlap with the previous and next subtrack. The default $i -1$ returns all subtracks. Can be negative, which
	means that space will be left between subtracks.

## Details

The output is always a single *tracks* object, which is convenient for many common analyses. If subtracks are to be considered separately for each track, use the function staggered together with lapply. Subtrack extraction always starts at the first position of the input track.

## Value

A *tracks* object is returned which contains all the subtracks of any track in the input *tracks* object that consist of exactly 'i' segments and overlap adjacent subtracks in 'overlap' segments.

## See Also

prefixes to extract all subtracks of a given length starting from the first coordinate in each track, subtracksByTime to extract all subtracks of a given length starting at some fixed timepoint, and selectSteps to extract single steps starting at a fixed timepoint from a subset of trackids.

subtracksByTime Extract Subtracks Starting at a Specific Time

## Description

Obtain all subtracks of i steps (i+1 positions) starting at a given timepoint t.

## Usage

subtracksByTime(X, t, i = 1, epsilon = 1e-04, tlo = t, thi = t)

## Arguments

Х	Tracks object to obtain subtracks from.
t	Timepoint at which the subtracks should start. This value is ignored if the and this are specified, see below.
i	Subtrack length (in number of steps). Set this to NULL to obtain subtracks of varying length but within a specified interval [tlo, thi] (see below).
epsilon	Small error allowed when comparing timepoints because of numerical inaccura- cies, see details. Timepoints in tracks are returned if they are within [tlo-epsilon, thi+epsilon].
tlo, thi	Interval specifying the timepoints to be returned. By default, these are not used and tracks starting at timepoint t with exactly i steps are returned; see details.

## Details

If i is specified, the given t is retrieved for all tracks in X that contain that timepoint, and any subtracks starting from that time that have exactly i steps are returned. For numerical reasons, timepoints in the data are allowed to deviate a small amount epsilon from t (because otherwise, equal timepoints can seem unequal because of very small deviations).

If i is set to NULL, subtracks are returned with all timepoints lying in the interval [ tlo - epsilon, thi + epsilon ]. These subtracks do NOT have to be of equal length.

#### Value

A *tracks* object is returned which contains all the subtracks of any track in the input *tracks* object that consist of exactly 'i' segments and start at the given timepoint t, OR a *tracks* object with all the timepoints of any track in the input *tracks* object that are between the and thi.

## See Also

subtracks to extract all subtracks of a given length, prefixes to extract all subtracks of a given length starting from the first coordinate in each track, selectSteps to extract single steps starting at a fixed timepoint from a subset of trackids, and timePoints to return all timepoints occurring in the dataset.

## Examples

## Get all the single steps (i=1) starting at the third timepoint in the T cell tracks. subT <- subtracksByTime( TCells, timePoints(TCells)[3], 1 )</pre>

## These all have the same number of steps: sapply( subT, nrow )

## Or set i to NULL and return all subtracks within the five first timepoints: subT2 <- subtracksByTime( TCells, NULL, i = NULL, tlo = timePoints( TCells )[1], thi = timePoints( TCells )[5] )

## These are not all the same length: sapply( subT2, nrow )

TCells

*Two-Photon Data: T Cells in a Lymph Node* 

#### Description

Labelled T cells were adoptively transfered and intravitally imaged (using two-photon microscopy) inside a peripheral lymph node of the recipient mouse. These data represent the characteristic "random-walk-like" motion pattern of T cells in lymph nodes.

#### Usage

data("TCells")

52

## timePoints

#### Format

An S3 object of class "tracks"; a list with 22 elements. Each element name identifies a cell track. Each element is a matrix containing the following four columns.

- t the time (in seconds)
- x The X coordinate (in micrometers)
- y The Y coordinate (in micrometers)
- z The Z coordinate (in micrometers)

#### Source

Data were generated in 2012 in the Mark J. Miller Lab, Department of Medicine, Washington University in St Louis, USA.

#### References

Zinselmeyer BH, Dempster J, Wokosin DL, Cannon JJ, Pless R, Parker I and Miller MJ (2009), Two-photon microscopy and multi-dimensional analysis of cell dynamics. *Methods in Enzymology*, **461**:349–78. doi:10.1016/S0076-6879(09)05416-0

Konjufca V and Miller MJ (2009), Imaging Listeria monocytogenes infection in vivo. *Current Topics in Microbiology and Immunology*, **334**:199–226. doi:10.1007/978-3-540-93864-4\_9

Kreisel D, Nava RG, Li W, Zinselmeyer BH, Wang B, Lai J, Pless R, Gelman AE, Krupnick AS, and Miller MJ (2010), In vivo two-photon imaging reveals monocyte-dependent neutrophil extravasation during pulmonary inflammation. *PNAS*, **107**(42):18073–18078. doi:10.1073/pnas.1008737107

## Examples

```
## load the tracks
data(TCells)
## visualize the tracks (calls function plot.tracks)
plot(TCells)
```

timePoints

Find All Unique Time Points in a Track Dataset

## Description

Return a vector of all the timepoints t found in any of the matrices in the tracks object.

#### Usage

timePoints(X)

#### Arguments

X a tracks object.

#### Value

A numeric vector of unique timepoints.

#### Examples

```
## Get all timepoints in the T cell dataset
tp <- timePoints( TCells )</pre>
```

timeStep

#### Compute Time Step of Tracks

## Description

Applies a summary statistics on the time intervals between pairs of consecutive positions in a track dataset.

## Usage

timeStep(x, FUN = median, na.rm = FALSE)

#### Arguments

Х	the input tracks.
FUN	the summary statistic to be applied.
na.rm	logical, indicates whether to remove missing values before applying FUN.

#### Details

Most track quantification depends on the assumption that track positions are recorded at constant time intervals. If this is not the case, then most of the statistics in this package (except for some very simple ones like duration) will not work. In reality, at least small fluctuations of the time steps can be expected. This function provides a means for quality control with respect to the tracking time.

#### Value

Summary statistic of the time intervals between two consecutive positions in a track dataset.

#### Examples

```
## Show tracking time fluctuations for the T cell data
d <- timeStep( TCells )
plot( sapply( subtracks( TCells, 1 ) , duration ) - d, ylim=c(-d,d) )</pre>
```

54

trackFeatureMap

## Description

Perform a quick dimensionality reduction visualization of a set of tracks according to a given vector of track measures.

## Usage

```
trackFeatureMap(
  tracks,
  measures,
  scale = TRUE,
  labels = NULL,
  method = "PCA",
  return.mapping = FALSE,
  ...
)
```

#### Arguments

tracks	the tracks that are to be clustered.
measures	a function, or a vector of functions (see TrackMeasures). Each function is expected to return a single number given a single track.
scale	logical indicating whether the measures values shall be scaled using the function scale before the mapping is performed.
labels	optional: a vector of labels of the same length as the track object. These are used to color points in the visualization.
method	"PCA" for a scatterplot along principal components, "MDS" for multidimensional scaling, "UMAP" for a UMAP. Note that for "UMAP", the uwot package must be installed.
return.mapping	logical: return the mapping object instead of only the plot? (defaults to ${\sf FALSE}$ ).
	additional parameters to be passed to the corresponding function: prcomp (for method="PCA"), cmdscale (for method="MDS"), or umap (for method="UMAP").

## Details

The measures are applied to each of the tracks in the given *tracks* object. According to the resulting values, the tracks are mapped to fewer dimensions using the chosen method. If scale is TRUE, the measure values are scaled to mean value 0 and standard deviation 1 (per measure) before the mapping.

The dimensionality reduction methods PCA, MDS, and UMAP each produce a scatterplot of all tracks as points, plotted along the principal component axes generated by the corresponding method.

#### Value

By default, only returns a plot. If return.clust=TRUE, also returns a clustering object as returned by hclust, kmeans, prcomp (returns \$x), cmdscale, or umap (returns \$layout). See the documentation of those functions for details on the output object.

#### See Also

getFeatureMatrix to obtain a feature matrix that can be used for manual clustering and plotting, and clusterTracks to perform hierarchical or k-means clustering on a tracks dataset.

#### Examples

## Map tracks according to speed, mean turning angle, straightness, and asphericity
## using multidimensional scaling, and store output.

```
cells <- c(TCells,Neutrophils)
real.celltype <- rep(c("T","N"),c(length(TCells),length(Neutrophils)))
## Prefix each track ID with its cell class to evaluate the clustering visually
names(cells) <- paste0(real.celltype,seq_along(cells))
map <- trackFeatureMap( cells, c(speed,meanTurningAngle,straightness, asphericity),
method = "MDS", return.mapping = TRUE )</pre>
```

TrackMeasures Track Measures

#### Description

Statistics that can be used to quantify tracks. All of these functions take a single track as input and give a single number as output.

#### Usage

```
trackLength(x)
duration(x)
speed(x)
displacement(x, from = 1, to = nrow(x))
squareDisplacement(x, from = 1, to = nrow(x))
displacementVector(x)
maxDisplacement(x)
displacementRatio(x)
```

```
outreachRatio(x)
straightness(x)
overallAngle(x, from = 1, to = nrow(x), xdiff = diff(x), degrees = FALSE)
meanTurningAngle(x, degrees = FALSE)
overallDot(x, from = 1, to = nrow(x), xdiff = diff(x))
overallNormDot(x, from = 1, to = nrow(x), xdiff = diff(x))
asphericity(x)
hurstExponent(x)
fractalDimension(x)
```

#### Arguments

x	a single input track; a matrix whose first column is time and whose remaining columns are a spatial coordinate.
from	index, or vector of indices, of the first row of the track.
to	index, or vector of indices, of last row of the track.
xdiff	row differences of x.
degrees	logical; should angles be returned in degrees rather than radians?

## Details

Some track measures consider only the first and last position (or steps) of a track, and are most useful in conjunction with aggregate.tracks; for instance, squareDisplacement combined with aggregate.tracks gives a mean square displacement plot, and overallAngle combined with aggregate.tracks gives a turning angle plot (see the examples for aggregate.tracks). To speed up computation of these measures on subtracks of the same track, the arguments from, to and possibly xdiff are exploited by aggregate.tracks.

#### Value

trackLength sums up the distances between subsequent position; in other words, it estimates the length of the underlying track by linear interpolation (usually an underestimation). The estimation could be improved in some circumstances by using interpolateTrack. The function returns a single, non-negative number.

duration returns the time elapsed between x's first and last positions (a single, non-negative number).

speed simply divides trackLength by duration

displacement returns the Euclidean distance between the track endpoints and squareDisplacement returns the squared Euclidean distance.

displacementVector returns the vector between the track endpoints. This vector has an element (can be negative) for each (x,y,z) dimension of the coordinates in the track.

maxDisplacement computes the maximal Euclidean distance of any position on the track from the first position.

displacementRatio divides the displacement by the maxDisplacement; outreachRatio divides the maxDisplacement by the trackLength (Mokhtari et al, 2013). Both measures return values between 0 and 1, where 1 means a perfectly straight track. If the track has trackLength 0, then NaN is returned.

straightness divides the displacement by the trackLength. This gives a number between 0 and 1, with 1 meaning a perfectly straight track. If the track has trackLength 0, then NaN is returned.

asphericity is a different appraoch to measure straightness (Mokhtari et al, 2013): it computes the asphericity of the set of positions on the track \_via\_ the length of its principal components. Again this gives a number between 0 and 1, with higher values indicating straighter tracks. Unlike straightness, however, asphericity ignores back-and-forth motion of the object, so something that bounces between two positions will have low straightness but high asphericity. We define the asphericity of every track with two or fewer positions to be 1. For one-dimensional tracks with one or more positions, NA is returned.

overallAngle Computes the angle (in radians) between the first and the last segment of the given track. Angles are measured symmetrically, thus the return values range from 0 to pi; for instance, both a 90 degrees left and right turns yield the value pi/2. This function is useful to generate autocorrelation plots (together with aggregate.tracks). Angles can also be returned in degrees, in that case: set degrees = TRUE.

meanTurningAngle averages the overallAngle over all adjacent segments of a given track; a low meanTurningAngle indicates high persistence of orientation, whereas for an uncorrelated random walk we expect 90 degrees. Note that angle measurements will yield NA values for tracks in which two subsequent positions are identical. By default returns angles in radians; use degrees = TRUE to return angles in degrees instead.

overallDot computes the dot product between the first and the last segment of the given track. This function is useful to generate autocovariance plots (together with aggregate.tracks).

overallNormDot computes the dot product between the unit vectors along the first and the last segment of the given track. This function is useful to generate autocorrelation plots (together with aggregate.tracks).

hurstExponent computes the corrected empirical Hurst exponent of the track. This uses the function hurstexp from the 'pracma' package. If the track has less than two positions, NA is returned. fractalDimension estimates the fractal dimension of a track using the function fd.estim.boxcount from the 'fractaldim' package. For self-affine processes in n dimensions, fractal dimension and Hurst exponent are related by the formula H = n + 1 - D. For non-Brownian motion, however, this relationship need not hold. Intuitively, while the Hurst exponent takes a global approach to the track's properties, fractal dimension is a local approach to the track's properties (Gneiting and Schlather, 2004).

#### tracks

## References

Zeinab Mokhtari, Franziska Mech, Carolin Zitzmann, Mike Hasenberg, Matthias Gunzer and Marc Thilo Figge (2013), Automated Characterization and Parameter–Free Classification of Cell Tracks Based on Local Migration Behavior. *PLoS ONE* **8**(12), e80808. doi:10.1371/journal.pone.0080808

Tillmann Gneiting and Martin Schlather (2004), Stochastic Models That Separate Fractal Dimension and the Hurst Effect. *SIAM Review* **46**(2), 269–282. doi:10.1137/S0036144501394387

## See Also

AngleAnalysis for methods to compute angles and distances between pairs of tracks, or of tracks to a reference point, direction, or plane.

#### Examples

```
## show a turning angle plot with error bars for the T cell data.
with( (aggregate(BCells,overallDot,FUN="mean.se",na.rm=TRUE)),{
    plot( mean ~ i, xlab="time step",
    ylab="turning angle (rad)", type="l" )
    segments( i, lower, y1=upper )
} )
```

tracks

Tracks Objects

#### Description

The function tracks is used to create tracks objects. as.tracks coerces its argument to a tracks object, and is.tracks tests for tracks objects. c can be used to combine (concatenate) tracks objects.

#### Usage

```
as.tracks(x, ...)
## S3 method for class 'list'
as.tracks(x, ...)
is.tracks(x)
## S3 method for class 'tracks'
c(...)
tracks(...)
```

#### Arguments

х	an object to be coerced or tested.
	for tracks, numeric matrices or objects that can be coerced to numeric matrices. Each matrix contains the data of one track. The first column is the time, and the remaining columns define a spatial position. Every given matrix has to contain the same number of columns, and at least two columns are necessary.
	For c, tracks objects to be combined. For as.tracks, further arguments passed to methods (currently not used).

## Details

Tracks objects are lists of matrices. Each matrix contains at least two columns; the first column is time, and the remaining columns are a spatial coordinate. The following naming conventions are used (and enforced by tracks): The time column has the name 't', and spatial coordinate columns have names 'x', 'y', 'z' if there are three or less coordinates, and 'x1',...,'xk' if there are  $k \ge 4$  coordinates. All tracks in an object must have the same number of dimensions. The positions in a track are expected to be sorted by time (and the constructor tracks enforces this).

#### Value

A tracks object.

#### Examples

```
## A single 1D track
x <- tracks( matrix(c(0, 8,
10, 9,
20, 7,
30, 7,
40, 6,
50, 5), ncol=2, byrow=TRUE ) )
## Three 3D tracks
x2 <- tracks( rbind(
    c(0,5,0), c(1,5,3), c(2,1,3), c(3,5,6) ),
    rbind( c(0,1,1),c(1,1,4),c(2,5,4),c(3,5,1),c(4,-3,1) ),
    rbind( c(0,7,0),c(1,7,2),c(2,7,4),c(3,7,7) ) )</pre>
```

vecAngle

Angle Between Two Vectors

#### Description

Compute the angle between two vectors a and b, which can be numeric vectors or matrices in which each row represents a numeric vector. In the last case, one angle is returned for each row. By default, angles are returned in degrees – set degrees = TRUE to return radians.

## wrapTrack

## Usage

vecAngle(a, b, degrees = TRUE)

#### Arguments

а	the first vector or set of vectors. Must be a numeric vector or a matrix where each row represents a numeric vector.
b	the second vector or set of vectors, for which angles with the vector (set) a must be computed. Must have the same dimensions as a.
degrees	logical: if TRUE (default), return angles in degrees instead of radians.

## Value

A single angle (if a and b are single vectors) or a numeric vector of angles (if a and b are matrices; in that case, the output vector contains one angle for each row in matrices a and b).

## Examples

```
## The angle between the vectors [0,1] and [1,0] is 90 degrees:
vecAngle( c(0,1), c(1,0) )
## The same holds for 3D angles:
vecAngle( c(0,1,0), c(1,0,0) )
```

wrapTrack Create Track Object from Single Track

## Description

Makes a tracks object containing the given track.

## Usage

```
wrapTrack(x)
```

## Arguments ×

the input track.

## Value

A list of class tracks containing only the input track x, which is assigned the name "1".

# Index

\*Topic **cluster** celltrackR, 24 \*Topic datasets BCells, 19 Neutrophils, 36 TCells, 52 \*Topic **spatial** celltrackR, 24 aggregate.tracks, 3, 25, 57, 58 analyzeCellPairs, 6, 7, 8 analyzeStepPairs, 6, 7, 8 AngleAnalysis, 8, 10-14, 25, 29-31, 59 angleCells, 6, 8, 10, 29 angleSteps, 7, 8, 10, 29 angleToDir, 8, 11 angleToPlane, 8, 12, 30 angleToPoint, 8, 14, 31 applyStaggered, 15, 25, 48 approx, 34, 35 as.data.frame.tracks, 16 as.list.tracks, 17, 24 as.tracks(tracks), 59 as.tracks.data.frame, 18 asphericity (TrackMeasures), 56

BCells, 19, 24 beaucheminTrack, 20, 25 bootstrapTrack, 21 boundingBox, 8, 22 brownianTrack, 23, 25

c.tracks(tracks), 59 cellPairs, 6, 8, 24 celltrackR, 24 cheatsheet, 24, 26, 26 clusterTracks, 27, 33, 56 cmdscale, 55, 56

displacement (TrackMeasures), 56

displacementRatio (TrackMeasures), 56 displacementVector (TrackMeasures), 56 distanceCells, 6, 8, 10, 28 distanceSteps, 7, 8, 11, 29 distanceToPlane, 8, 13, 30 distanceToPoint, 8, 14, 31 duration, 54, 57 duration (TrackMeasures), 56

fd.estim.boxcount, 58
filterTracks, 32
fractalDimension (TrackMeasures), 56

getFeatureMatrix, 28, 32, 56

hclust, 27, 28, 56 hotellingsTest, 33 hurstexp, 58 hurstExponent (TrackMeasures), 56

interpolateTrack, 34, 44, 57
is.tracks(tracks), 59

kmeans, 27, 28, 56

list, 24

match.fun, 4
maxDisplacement(TrackMeasures), 56
maxTrackLength, 35
meanTurningAngle(TrackMeasures), 56

Neutrophils, 24, 36 normalizeToDuration, 37 normalizeTracks, 37

outreachRatio (TrackMeasures), 56 overallAngle, 15 overallAngle (TrackMeasures), 56 overallDot (TrackMeasures), 56 overallNormDot (TrackMeasures), 56

## INDEX

plot, 38, 40 plot.tracks, 24, 38 plot3d, 39, 39 plotTrackMeasures, 39 points, 38, 40 prcomp, 55, 56 prefixes, 41, 45, 51, 52 projectDimensions, 22, 42 read.table, 43 read.tracks.csv, 42 repairGaps, 44 replicate, 46 scale, 27, 55 scatterplot3d, 39 selectSteps, 41, 45, 51, 52 selectTracks, 46 simulateTracks, 46 sort.tracks, 24, 47 speed, 25 speed (TrackMeasures), 56 spline, 35 splitTrack, 48 squareDisplacement (TrackMeasures), 56 staggered, 48, 51 stepPairs, 7, 8, 49 straightness, 25, 58 straightness (TrackMeasures), 56 subsample, 50 subtracks, 25, 41, 45, 50, 52 subtracksByTime, 41, 45, 51, 51 TCells, 24, 52 timePoints, 11, 29, 45, 52, 53 timeStep, 22, 24, 44, 54 trackFeatureMap, 28, 33, 55 trackLength, 57 trackLength (TrackMeasures), 56 TrackMeasures, 9, 25, 27, 32, 37, 40, 46, 48, 55, 56 tracks, 24, 44, 46, 50, 59 umap, 55, 56 vecAngle, 60 wrapTrack, 4, 61