# Package 'rearrr'

August 28, 2022

```
Title Rearranging Data
Version 0.3.1
Description Arrange data by a set of methods. Use rearrangers to reorder
     data points and mutators to change their values. From basic utilities,
     to centering the greatest value, to swirling in 3-dimensional space,
     'rearrr' enables creativity when plotting and experimenting with data.
License MIT + file LICENSE
URL https://github.com/ludvigolsen/rearrr
BugReports https://github.com/ludvigolsen/rearrr/issues
Depends R (>= 3.5)
Imports checkmate (>= 2.0.0),
     dplyr (>= 0.8.5),
     lifecycle,
     plyr,
     purrr (>= 0.3.4),
     rlang (>= 0.4.7),
     R6,
     stats,
     tibble,
     utils
Suggests covr,
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     knitr,
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124

Index

apply_transformation_matrix	5
center_max	7
center_min	9
centroid	11
circularize	12
closest_to	
cluster_groups	
create_dimming_fn	
create_n_fn	
create_origin_fn	
degrees_to_radians	
dim_values	
distance	
expand_distances	
expand_distances_each	
FixedGroupsPipeline	
FixedGroupsTransformation	
flip_values	
furthest_from	
GeneratedPipeline	
Generated Transformation	
generate_clusters	
hexagonalize	
is_most_centered	
median_index	
midrange	
min_max_scale	
most_centered	
pair_extremes	
position_max	
position_min	
rev_windows	
roll_elements	
roll values	
rotate 2d	
rotate 3d	87
shear 2d	
shear 3d	
shuffle_hierarchy	
square	
swirl_2d	
swirl_3d	
to_unit_length	
transfer_centroids	
Transformation	
triangularize	
triplet_extremes	
vector_length	
-	

angle 3

angle

Calculate the angle to an origin

## Description

## [Experimental]

Calculates the angle between each data point (x2, y2) and the origin (x1, y1) with:

$$atan2(y2-y1, x2-x1)$$

And converts to degrees [0-360), measured counterclockwise from the  $\{x > x1, y = y1\}$  line.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and finding the angle to e.g. the centroid of each group.

## Usage

```
angle(
  data,
  x_col = NULL,
  y_col = NULL,
  origin = NULL,
  origin_fn = NULL,
  degrees_col_name = ".degrees",
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

## Arguments

overwrite

data	data.frame or vector.	
x_col	Name of x column in `data`. If NULL and `data` is a vector, the index of `data` is used. If `data` is a data. frame, it must be specified.	
y_col	Name of y column in `data`. If `data` is a data.frame, it must be specified.	
origin	Coordinates of the origin to calculate angle to. A scalar to use in all dimensions or a vector with one scalar per dimension.	
	<b>N.B.</b> Ignored when `origin_fn` is not NULL.	
origin_fn	Function for finding the origin coordinates.	
	<b>Input</b> : Each column will be passed as a vector in the order of `cols`.	
	Output: A vector with one scalar per dimension.	
	Can be created with create_origin_fn() if you want to apply the same function to each dimension.	
	E.g. `create_origin_fn(median)` would find the median of each column.	
	<pre>Built-in functions are centroid(), most_centered(), and midrange()</pre>	
degrees_col_name		
	Name of new column with the degrees.	
origin_col_name		

Whether to allow overwriting of existing columns. (Logical)

Name of new column with the origin coordinates. If NULL, no column is added.

4 angle

#### Value

data.frame (tibble) with the additional columns (degrees and origin coordinates).

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

Other measuring functions: distance(), vector\_length()

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  x'' = runif(20),
  "y" = runif(20),
  "g" = rep(1:4, each = 5)
# Calculate angles in the two dimensions (x and y)
# With the origin at x=0.5, y=0.5
df_angles <- angle(</pre>
  data = df,
  x_{col} = "x"
  y_{col} = "y",
  origin = c(0.5, 0.5)
df_angles
# Plot points with degrees
# Degrees are measured counterclockwise around the
# positive side of the x-axis
df_angles %>%
  ggplot(aes(x = x, y = y, color = .degrees)) +
  geom\_segment(aes(x = 0.5, xend = 1, y = 0.5, yend = 0.5), color = "magenta") +
  geom_point() +
  theme_minimal()
# Calculate angles to the centroid for each group in 'g'
angle(
  data = dplyr::group_by(df, g),
  x_{col} = "x",
  y_col = "y",
  origin_fn = centroid
```

```
apply_transformation_matrix
```

Apply transformation matrix to a set of columns

## **Description**

## [Experimental]

Perform matrix multiplication with a transformation matrix and a set of data. frame columns.

The data points in `data` are moved prior to the transformation, to bring the origin to 0 in all dimensions. After the transformation, the inverse move is applied to bring the origin back to its original position. See `Details` section.

The columns in `data` are transposed, making the operation (without the origin movement):

$$matdata[,cols]^T$$

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and transforming around e.g. the centroid of each group.

## Usage

```
apply_transformation_matrix(
  data,
  mat,
  cols,
  origin = NULL,
  origin_fn = NULL,
  suffix = "_transformed",
  keep_original = TRUE,
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

## Arguments

data	data.frame or vector.
mat	Transformation matrix. Must have the same number of columns as `cols`.
cols	Columns to mutate values of. Must be specified when `data` is a data.frame.
origin	Coordinates of the origin. Vector with the same number of elements as `cols` (i.e. origin_x, origin_y,). Ignored when `origin_fn` is not NULL.
origin_fn	Function for finding the origin coordinates.
	<b>Input</b> : Each column will be passed as a vector in the order of `cols`.
	Output: A vector with one scalar per dimension.
	Can be created with create_origin_fn() if you want to apply the same function to each dimension.
	E.g. `create_origin_fn(median)` would find the median of each column.
	<pre>Built-in functions are centroid(), most_centered(), and midrange()</pre>

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### **Details**

Example with 2 columns (x, y) and a 2x2 transformation matrix:

```
• Move origin to (0, 0):
```

```
x = x - origin_x
```

$$y = y - origin_y$$

• Convert to transposed matrix:

```
data_mat = rbind(x, y)
```

• Matrix multiplication:

```
transformed = mat %*% data_mat
```

• Move origin to original position (after extraction from transformed):

```
x = x + origin_x
```

## y = y + origin\_y

#### Value

data.frame (tibble) with the new, transformed columns and the origin coordinates.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

```
Other mutate functions: cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)

# Set seed
set.seed(3)

# Create a data frame
df <- data.frame(
   "x" = 1:12,
   "y" = 13:24,</pre>
```

center\_max 7

```
z" = runif(12),
  g'' = c(
   1, 1, 1, 1, 2, 2,
   2, 2, 3, 3, 3, 3
# Apply identity matrix
mat <- matrix(c(1, 0, 0, 0, 1, 0, 0, 0, 1), nrow = 3)
apply_transformation_matrix(
 data = df,
 mat = mat,
 cols = c("x", "y", "z"),
 origin = c(0, 0, 0)
# Apply rotation matrix
# 90 degrees around z-axis
# Origin is the most centered point
mat <- matrix(c(0, 1, 0, -1, 0, 0, 0, 0, 1), nrow = 3)
res <- apply_transformation_matrix(</pre>
 data = df,
 mat = mat,
 cols = c("x", "y", "z"),
 origin_fn = most_centered
# Plot the rotation
# z wasn't changed so we plot x and y
res %>%
  ggplot(aes(x = x, y = y)) +
  geom_point() +
  geom\_point(aes(x = x\_transformed, y = y\_transformed)) +
  theme_minimal()
# Apply rotation matrix to grouped data frame
# Around centroids
# Same matrix as before
res <- apply_transformation_matrix(</pre>
 data = dplyr::group_by(df, g),
 mat = mat,
 cols = c("x", "y", "z"),
 origin_fn = centroid
# Plot the rotation
res %>%
  ggplot(aes(x = x, y = y, color = g)) +
  geom_point() +
  geom\_point(aes(x = x\_transformed, y = y\_transformed)) +
  theme_minimal()
```

8 center\_max

## **Description**

## [Experimental]

The highest value is positioned in the middle with the other values decreasing around it.

#### Example:

```
The column values:
```

```
c(1, 2, 3, 4, 5)
```

#### are ordered as:

```
c(1, 3, 5, 4, 2)
```

## Usage

```
center_max(data, col = NULL, shuffle_sides = FALSE)
```

#### **Arguments**

data data.frame or vector.

col Column to create sorting factor by. When `NULL` and `data` is a data. frame,

the row numbers are used.

shuffle\_sides Whether to shuffle which elements are left and right of the center. (Logical)

## Value

The sorted data.frame (tibble) / vector.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_min(), closest_to(), furthest_from(), pair_extremes(), position_max(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)

# Set seed
set.seed(1)

# Create a data frame
df <- data.frame(
   "index" = 1:10,
   "A" = sample(1:10),
   "B" = runif(10),
   "C" = LETTERS[1:10],
   "G" = c(
    1, 1, 1, 2, 2,
    2, 3, 3, 3, 3</pre>
```

center\_min 9

```
),
  stringsAsFactors = FALSE
# Center by the index (row numbers)
center_max(df)
# Center by each of the columns
center_max(df, col = "A")
center_max(df, col = "B")
center_max(df, col = "C")
# Randomize which elements are left and right of the center
center_max(df, col = "A", shuffle_sides = TRUE)
# Grouped by G
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  center_max(col = "A")
# Plot the centered values
plot(x = 1:10, y = center_max(df, col = "B")$B)
plot(x = 1:10, y = center_max(df, col = "B", shuffle_sides = TRUE)$B)
```

center\_min

Centers the lowest value with values increasing around it

## **Description**

#### [Experimental]

The lowest value is positioned in the middle with the other values increasing around it.

#### Example:

The column values:

```
c(1, 2, 3, 4, 5)
```

#### are ordered as:

```
c(5, 3, 1, 2, 4)
```

#### Usage

```
center_min(data, col = NULL, shuffle_sides = FALSE)
```

## Arguments

data data.frame or vector.

col Column to create sorting factor by. When `NULL` and `data` is a data.frame,

the row numbers are used.

shuffle\_sides Whether to shuffle which elements are left and right of the center. (Logical)

#### Value

The sorted data.frame (tibble) / vector.

10 center\_min

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_max(), closest_to(), furthest_from(), pair_extremes(), position_max(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "index" = 1:10,
  "A" = sample(1:10),
  "B" = runif(10),
  "C" = LETTERS[1:10],
  G'' = c(
   1, 1, 1, 2, 2,
   2, 3, 3, 3, 3
 ),
  stringsAsFactors = FALSE
# Center by the index (row numbers)
center_min(df)
# Center by each of the columns
center_min(df, col = "A")
center_min(df, col = "B")
center_min(df, col = "C")
# Randomize which elements are left and right of the center
center_min(df, col = "A", shuffle_sides = TRUE)
# Grouped by G
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  center_min(col = "A")
# Plot the centered values
plot(x = 1:10, y = center\_min(df, col = "B")$B)
plot(x = 1:10, y = center\_min(df, col = "B", shuffle\_sides = TRUE)$B)
```

centroid 11

centroid

Find the coordinates for the centroid

## **Description**

## [Experimental]

Calculates the mean of each passed vector/column.

## Usage

```
centroid(..., cols = NULL, na.rm = FALSE)
```

## **Arguments**

... Numeric vectors or a single data.frame.

cols Names of columns to use when `...` is a single data.frame.

na.rm Whether to ignore missing values when calculating means. (Logical)

#### Value

Means of the supplied vectors/columns. Either as a vector or a data.frame.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

 $Other coordinate functions: create\_origin\_fn(), is\_most\_centered(), midrange(), most\_centered()$ 

```
# Attach packages
library(rearrr)
library(dplyr)

# Set seed
set.seed(1)

# Create three vectors
x <- runif(10)
y <- runif(10)
z <- runif(10)

# Find centroid coordinates
# Aka. the means of each vector
centroid(x, y, z)

#
# For data.frames
#
# Create data frame</pre>
```

12 circularize

```
df <- data.frame(
  "x" = x,
  "y" = y,
  "z" = z,
  "g" = rep(1:2, each = 5)
)

# Find centroid coordinates
# Aka. the means of each column
centroid(df, cols = c("x", "y", "z"))

# When 'df' is grouped
df %>%
  dplyr::group_by(g) %>%
  centroid(cols = c("x", "y", "z"))
```

circularize

Create x-coordinates so the points form a circle

## Description

## [Experimental]

Create the x-coordinates for a vector of y-coordinates such that they form a circle.

This will likely look most like a circle when the y-coordinates are somewhat equally distributed, e.g. a uniform distribution.

## Usage

```
circularize(
  data,
  y_col = NULL,
  .min = NULL,
  .max = NULL,
  offset_x = 0,
  keep_original = TRUE,
  x_col_name = ".circle_x",
  degrees_col_name = ".degrees",
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

## **Arguments**

data	data.frame or vector.
y_col	Name of column in `data` with y-coordinates to create x-coordinates for.
.min	Minimum y-coordinate. If NULL, it is inferred by the given y-coordinates.
.max	Maximum y-coordinate. If NULL, it is inferred by the given y-coordinates.
offset_x	Value to offset the x-coordinates by.
keep_original	Whether to keep the original columns. (Logical)
	Some columns may have been overwritten, in which case only the newest ver-
	sions are returned.

circularize 13

```
x_col_name Name of new column with the x-coordinates.

degrees_col_name

Name of new column with the angles in degrees. If NULL, no column is added.

Angling is counterclockwise around (0, 0) and starts at (max(x), 0).

origin_col_name

Name of new column with the origin coordinates (center of circle). If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)
```

#### Value

data.frame (tibble) with the added x-coordinates and the angle in degrees.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

Other forming functions: hexagonalize(), square(), triangularize()

```
# Attach packages
library(rearrr)
library(dplyr)
library(purrr)
library(ggplot2)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
 y'' = runif(200),
  g'' = factor(rep(1:5, each = 40))
# Circularize 'y'
df_circ <- circularize(df, y_col = "y")</pre>
df_circ
# Plot circle
df_circ %>%
  ggplot(aes(x = .circle_x, y = y, color = .degrees)) +
  geom_point() +
  theme_minimal()
#
# Grouped circularization
# Circularize 'y' for each group
# First cluster the groups a bit to move the
# circles away from each other
```

14 circularize

```
df_circ <- df %>%
  cluster_groups(
   cols = "y",
    group_cols = "g",
   suffix = "",
   overwrite = TRUE
  ) %>%
  dplyr::group_by(g) %>%
  circularize(
   y_{col} = "y",
   overwrite = TRUE
  )
# Plot circles
df_circ %>%
 ggplot(aes(x = .circle_x, y = y, color = g)) +
  geom_point() +
 theme_minimal()
# Specifying minimum value
# Specify minimum value manually
df_circ <- circularize(df, y_col = "y", .min = -2)</pre>
df\_circ
# Plot circle
df circ %>%
 ggplot(aes(x = .circle_x, y = y, color = .degrees)) +
 geom_point() +
 theme_minimal()
# Multiple circles by contraction
# Start by circularizing 'y'
df_circ <- circularize(df, y_col = "y")</pre>
# Contract '.circle_x' and 'y' towards the centroid
# To contract with multiple multipliers at once,
# we wrap the call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  .x = 1:10 / 10,
  .f = function(mult) {
    expand_distances(
      data = df_circ,
      cols = c(".circle_x", "y"),
      multiplier = mult,
      origin_fn = centroid,
      overwrite = TRUE
   )
 }
)
df_expanded
```

closest\_to 15

```
df_expanded %>%
  ggplot(aes(
    x = .circle_x_expanded, y = y_expanded,
    color = .degrees, alpha = .multiplier
  )) +
  geom_point() +
  theme_minimal()
```

closest\_to

Orders values by shortest distance to an origin

## **Description**

#### [Experimental]

Values are ordered by how close they are to the origin.

In 1d (when `cols` has length 1), the origin can be thought of as a target value. In n dimensions, the origin can be thought of as coordinates.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data.frame and ordering the rows by their distance to the centroid of each group.

The \*\_vec() version takes and returns a vector.

#### Example:

```
The column values:
```

```
c(1, 2, 3, 4, 5)
and origin = 2
are ordered as:
c(2, 1, 3, 4, 5)
```

## Usage

```
closest_to(
  data,
  cols = NULL,
  origin = NULL,
  origin_fn = NULL,
  shuffle_ties = FALSE,
  origin_col_name = ".origin",
  distance_col_name = ".distance",
  overwrite = FALSE
)
closest_to_vec(data, origin = NULL, origin_fn = NULL, shuffle_ties = FALSE)
```

## Arguments

```
data data.frame or vector.
```

cols Column(s) to create sorting factor by. When `NULL` and `data` is a data. frame,

the row numbers are used.

16 closest\_to

origin Coordinates of the origin to calculate distances to. A scalar to use in all dimensions or a vector with one scalar per dimension.

**N.B.** Ignored when `origin\_fn` is not `NULL`.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

 $E.g. \ \verb|`create_origin_fn(median)| \ would find the median of each column.$ 

Built-in functions are centroid(), most\_centered(), and midrange()

shuffle\_ties
origin\_col\_name

Whether to shuffle elements with the same distance to the origin. (Logical)

Name of new column with the origin coordinates. If `NULL`, no column is added.

distance\_col\_name

Name of new column with the distances to the origin. If `NULL`, no column is

added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### Value

The sorted data.frame (tibble) / vector.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

```
Other rearrange functions: center_max(), center_min(), furthest_from(), pair_extremes(), position_max(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

Other distance functions: dim\_values(), distance(), expand\_distances\_each(), expand\_distances(), furthest\_from(), swirl\_2d(), swirl\_3d()

```
# Attach packages
library(rearrr)
library(dplyr)

# Set seed
set.seed(1)

# Create a data frame
df <- data.frame(
   "index" = 1:10,
   "A" = sample(1:10),
   "B" = runif(10),
   "G" = c(
    1, 1, 1, 2, 2,
    2, 3, 3, 3, 3</pre>
```

closest\_to 17

```
),
 stringsAsFactors = FALSE
# Closest to 3 in a vector
closest_to_vec(1:10, origin = 3)
# Closest to the third row (index of data.frame)
closest_to(df, origin = 3)$index
# By each of the columns
closest_to(df, cols = "A", origin = 3)$A
closest_to(df, cols = "A", origin_fn = most_centered)$A
closest_to(df, cols = "B", origin = 0.5)$B
closest_to(df, cols = "B", origin_fn = centroid)$B
# Shuffle the elements with the same distance to the origin
closest_to(df,
 cols = "A",
 origin_fn = create_origin_fn(median),
  shuffle_ties = TRUE
)$A
# Grouped by G
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  closest_to(
   cols = "A"
    origin_fn = create_origin_fn(median)
# Plot the rearranged values
plot(
  x = 1:10,
  y = closest_to(df,
   cols = "B",
   origin_fn = create_origin_fn(median)
  )$B,
  xlab = "Position",
 ylab = "B"
)
plot(
 x = 1:10,
  y = closest_to(df,
   cols = "A",
   origin_fn = create_origin_fn(median),
   shuffle_ties = TRUE
  )$A,
 xlab = "Position",
 ylab = "A"
# In multiple dimensions
df %>%
  closest_to(cols = c("A", "B"), origin_fn = most_centered)
```

18 cluster\_groups

cluster\_groups

Move data points into clusters

#### **Description**

#### [Experimental]

Transform values such that the elements in each group move closer to their centroid.

## Usage

```
cluster_groups(
 data,
  cols,
  group_cols = NULL,
  scale_min_fn = function(x) {
     quantile(x, 0.025)
 },
  scale_max_fn = function(x) {
     quantile(x, 0.975)
 },
  keep_centroids = FALSE,
 multiplier = 0.05,
  suffix = "_clustered",
 keep_original = TRUE,
 overwrite = FALSE
)
```

## Arguments

data data.frame. If `group\_cols` is NULL, it must be grouped with dplyr::group\_by().

cols Names of columns in `data` to mutate. Each column is considered a dimension

to contract distances in.

group\_cols Names of grouping columns in `data`. Must be distinct from the names in

`cols`.

If NULL and `data` is grouped, those groups are used instead.

scale\_min\_fn, scale\_max\_fn

Function to find the minimum/maximum value in the original data when rescal-

ing the contracted data.

Input: A numeric vector.
Output: A numeric scalar.

keep\_centroids Whether to ensure the clusters have their original centroids. (Logical)

multiplier Numeric constant to multiply the distance to the group centroid by. A smaller

value makes the clusters more compact and vice versa.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

overwrite Whether to allow overwriting of existing columns. (Logical)

cluster\_groups 19

#### **Details**

- Contracts the distance from each data point to the centroid of its group.
- Performs MinMax scaling such that the scale of the data points is *similar* to the original data.
- If enabled (not default), the centroids are moved to the original centroids.

#### Value

```
data.frame (tibble) with the clustered columns.
```

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other mutate functions: apply_transformation_matrix(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()

Other clustering functions: generate_clusters(), transfer_centroids()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Set seed
set.seed(2)
# Create a data frame
df <- data.frame(</pre>
  x'' = runif(50),
  "y" = runif(50),
  z'' = runif(50),
  "g" = rep(c(1, 2, 3, 4, 5), each = 10)
# Move the data points into clusters
cluster_groups(df,
 cols = c("x", "y"),
  group_col = "g"
cluster_groups(df,
  cols = c("x", "y"),
  group_col = "g",
 multiplier = 0.1
cluster_groups(df,
  cols = c("x"),
  group_col = "g",
 multiplier = 0.1
#
```

20 cluster\_groups

```
# Plotting clusters
\# Cluster x and y for each group in g
df_clustered <- cluster_groups(</pre>
  data = df,
 cols = c("x", "y"),
 group_col = "g"
# Plot the clusters over the original data points
# As we work with random data, the cluster might overlap
  df_clustered,
 aes(x = x\_clustered, y = y\_clustered, color = factor(g))
) +
  # Original data
  geom_point(aes(x = x, y = y), alpha = 0.3, size = 0.8) +
  # Clustered data
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "g")
# Maintain original group centroids
df_clustered <- cluster_groups(</pre>
 data = df,
 cols = c("x", "y"),
 group_col = "g",
 keep_centroids = TRUE
# Plot the clusters over the original data points
# As we work with random data, the cluster might overlap
ggplot(
  df_clustered,
 aes(x = x\_clustered, y = y\_clustered, color = factor(g))
  # Original data
  geom_point(aes(x = x, y = y), alpha = 0.3, size = 0.8) +
  # Clustered data
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "g")
# Three dimensions
# Cluster in 3d
df_clustered <- cluster_groups(</pre>
  data = df,
 cols = c("x", "y", "z"),
 group_col = "g"
)
```

create\_dimming\_fn 21

```
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
    x = df_clustered$x_clustered,
    y = df_clustered$y_clustered,
    z = df_clustered$z_clustered,
    type = "scatter3d",
    mode = "markers",
    color = df_clustered$g
)
## End(Not run)
```

create\_dimming\_fn

Create dimming\_fn function

## **Description**

## [Experimental]

Creates a function that takes 2 inputs (`x`, `d`) and performs the operation:

```
x * (numerator/((add_to_distance + d)^e x ponent))
```

Here, `x` is the current value and `d` is its distance to an origin. The greater the distance, the more we will dim the value of `x`.

With the default values, the returned function is:

```
function(x, d){
  x * (1 / ((1 + d) ^ 2))
}
```

## Usage

```
create_dimming_fn(numerator = 1, exponent = 2, add_to_distance = 1)
```

## **Arguments**

numerator The numerator. Defaults to 1. exponent The exponent. Defaults to 2. add\_to\_distance

Constant to add to the distance before exponentiation. Ensures dimming even when the distance (d) is below 1. Defaults to 1.

#### Value

Function with the arguments x and d, with both expected to be numeric vectors. More specifically:

```
function(x, d){
   x * (numerator / ((add_to_distance + d) ^ exponent))
}
```

22 create\_n\_fn

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

Other function creators: create\_n\_fn(), create\_origin\_fn()

## **Examples**

```
# Attach packages
library(rearrr)
library(ggplot2)
# Set seed
set.seed(1)
# Create two vectors
x <- runif(10)</pre>
d <- runif(10, max = 0.5)
\# Create dimming_fn with an add_to_distance of 0
# Note: In practice this risks zero-division
non_smoothed_dimming_fn <- create_dimming_fn(add_to_distance = 0)</pre>
non_smoothed_dimming_fn
as.list(environment(non_smoothed_dimming_fn))
# Use median_origin_fn
non_smoothed_dimming_fn(x, d)
# Plotting the dimming
# Create data.frame with distance-based dimming
df <- data.frame(</pre>
  "x" = 1,
  "d" = 1:10
df$x_dimmed <- non_smoothed_dimming_fn(df$x, df$d)</pre>
# Plot the dimming
ggplot(df, aes(x=d, y=x_dimmed)) +
  geom_point() +
  geom_line() +
  theme_minimal()
```

create\_n\_fn

Create n\_fn function

## Description

## [Experimental]

Creates a function that applies a supplied function to all input vectors, or their indices, and rounds the results

As used with roll\_elements(). E.g. to find the median index in a subset of a grouped data.frame.

create\_n\_fn 23

#### Usage

```
create_n_fn(fn, use_index = FALSE, negate = FALSE, round_fn = round, ...)
```

## **Arguments**

fn Function to apply to each dimension. Should return a numeric scalar.

use\_index Whether to apply `fn` to the indices of the vectors. (Logical)

The indices are created with seq\_along(x).

negate Whether to negate the result. I.e. to multiply it with -1. (Logical)

round\_fn Function for rounding results of `fn`.

Rounding is done prior to negation.

E.g. round, floor, or ceiling.

To avoid rounding, supply identity.

... Arguments for `fn`. E.g. `na.rm = TRUE`.

#### Value

Function with the dots (`...`) argument that applies the `fn` function to each element in `...` (or indices thereof) (usually one vector per dimension). The results are rounded with `round\_fn`.

Note: The dots argument in the generated function should not to be confused with the dots argument in create\_n\_fn()).

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other n functions: median_index()
Other function creators: create_dimming_fn(), create_origin_fn()
```

```
# Attach packages
library(rearrr)

# Set seed
set.seed(1)

# Create three vectors
x <- runif(10)
y <- runif(10)
z <- runif(10)

# Create n_fn that gets the median index
# and rounds down with floor()
median_index_fn <- create_n_fn(median, use_index = TRUE, round_fn = floor)

# Use median_index_fn
median_index_fn
median_index_fn(x, y, z)

# Create n_fn that gets the median of each dimension</pre>
```

24 create\_origin\_fn

```
median_n_fn <- create_n_fn(median)</pre>
# Use median_origin_fn
median_n_fn(x, y, z)
# Should be the same as
round(c(median(x), median(y), median(z)))
# Use mean and ignore missing values
mean_n_fn <- create_n_fn(mean, na.rm = TRUE)</pre>
# Add missing values
x[[2]] <- NA
y[[5]] <- NA
\# Use mean_n_fn
mean_nfn(x, y, z)
# Should be the same as
round(c(
 mean(x, na.rm = TRUE),
 mean(y, na.rm = TRUE),
 mean(z, na.rm = TRUE)
))
```

create\_origin\_fn

Create origin\_fn function

## Description

## [Experimental]

Creates a function that applies a supplied function to all input vectors.

## Usage

```
create_origin_fn(fn, ...)
```

## **Arguments**

```
fn Function to apply to each dimension. Should return a numeric scalar.

Arguments for `fn`. E.g. `na.rm = TRUE`.
```

## Value

Function with the dots (...) argument that applies the `fn` function to each element in ... (usually one vector per dimension).

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

degrees\_to\_radians 25

#### See Also

```
Other coordinate functions: centroid(), is_most_centered(), midrange(), most_centered()
Other function creators: create_dimming_fn(), create_n_fn()
```

## **Examples**

```
# Attach packages
library(rearrr)
# Set seed
set.seed(1)
# Create three vectors
x <- runif(10)
y <- runif(10)
z <- runif(10)</pre>
# Create origin_fn that gets the median of each dimension
median_origin_fn <- create_origin_fn(median)</pre>
# Use median_origin_fn
median_origin_fn(x, y, z)
# Should be the same as
c(median(x), median(y), median(z))
# Use mean and ignore missing values
mean_origin_fn <- create_origin_fn(mean, na.rm = TRUE)</pre>
# Add missing values
x[[2]] <- NA
y[[5]] <- NA
# Use mean_origin_fn
mean\_origin\_fn(x, y, z)
# Should be the same as
c(mean(x, na.rm = TRUE),
 mean(y, na.rm = TRUE),
 mean(z, na.rm = TRUE)
```

degrees\_to\_radians

Conversion between radians and degrees

## Description

## [Experimental]

Convert degrees to radians or radians to degrees.

#### Usage

```
degrees_to_radians(degrees)
radians_to_degrees(radians)
```

#### **Arguments**

degrees vector of degrees to convert to radians with

'degrees' \*  $(\pi/180)$ 

radians vector of radians to convert to degrees with

'radians'/ $(\pi/180)$ 

## Value

vector with converted degrees/radians.

Missing values (NAs) are returned as they are.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## **Examples**

```
# Attach packages
library(rearrr)
library(dplyr)

# Degrees to radians
degrees_to_radians(c(90, 180, 270))

# Radians to degrees
radians_to_degrees(c(pi / 2, pi, 1.5 * pi))

# Get back the original degrees
radians_to_degrees(degrees_to_radians(c(90, 180, 270)))
```

dim\_values

Dim values of a dimension based on the distance to an n-dimensional origin

## Description

## [Experimental]

Dims the values in the dimming dimension (last by default) based on the data point's distance to the origin.

Distance is calculated as:

$$d(P1, P2) = sqrt((x2 - x1)^2 + (y2 - y1)^2 + (z2 - z1)^2 + \dots)$$

The default 'dimming\_fn' multiplies by the inverse-square of 1 + distance and is calculated as:

$$dimming_f n(x, d) = x * (1/(1+d)^2)$$

Where x is the value in the dimming dimension. The +1 is added to ensure that values are dimmed even when the distance is below 1. The quickest way to change the exponent or the +1 is with  $create\_dimming\_fn()$ .

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and dimming around e.g. the centroid of each group.

## Usage

```
dim_values(
  data,
  cols,
  dimming_fn = create_dimming_fn(numerator = 1, exponent = 2, add_to_distance = 1),
  origin = NULL,
  origin_fn = NULL,
  dim_col = tail(cols, 1),
  suffix = "_dimmed",
  keep_original = TRUE,
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

## **Arguments**

data

data.frame or vector.

cols

Names of columns in `data` to calculate distances from. The dimming column (`dim\_col`) is dimmed based on all the columns. Each column is considered a dimension.

**N.B.** when the dimming dimension is included in `cols`, it is used in the distance calculation as well.

dimming\_fn

Function for calculating the dimmed values.

**Input**: Two (2) input arguments:

- 1. A numeric vector with the values in the dimming dimension.
- 2. A numeric vector with corresponding distances to the origin.

Output: A numeric vector with the same length as the input vectors.

E.g.:

```
function(x, d){
  x * (1 / ((1 + d) ^ 2))
}
```

This kind of dimming function can be created with create\_dimming\_fn(), which for instance makes it easy to change the exponent (the 2 above).

origin

Coordinates of the origin to dim around. A scalar to use in all dimensions or a vector with one scalar per dimension.

**N.B.** Ignored when `origin\_fn` is not NULL.

origin\_fn

Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

 $E.g. \verb|`create_origin_fn(median)`| would find the median of each column.$ 

**Built-in functions** are centroid(), most\_centered(), and midrange()

dim\_col Name of column to dim. Default is the last column in `cols`.

When the  $\injty$  is not present in  $\injty$  is not used in the distance

calculation.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### **Details**

• Calculates distances to origin with:

$$d(P1, P2) = sqrt((x2 - x1)^{2} + (y2 - y1)^{2} + (z2 - z1)^{2} + \dots)$$

• Applies the `dimming\_fn` to the `dim\_col` based on the distances.

#### Value

data.frame (tibble) with the dimmed column, along with the origin coordinates.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()
```

Other distance functions: closest\_to(), distance(), expand\_distances\_each(), expand\_distances(), furthest\_from(), swirl\_2d(), swirl\_3d()

```
# Attach packages
library(rearrr)
library(dplyr)
library(purrr)
library(ggplot2)
# Set seed
set.seed(7)
```

```
# Create a data frame with clusters
df <- generate_clusters(</pre>
  num_rows = 70,
  num\_cols = 3,
 num_clusters = 5,
 compactness = 1.6
  dplyr::rename(x = D1, y = D2, z = D3) \%%
  dplyr::mutate(o = 1)
# Dim the values in the z column
dim_values(
 data = df,
 cols = c("x", "y", "z"),
 origin = c(0.5, 0.5, 0.5)
# Dim the values in the `o` column (all 1s)
# around the centroid
dim_values(
 data = df,
 cols = c("x", "y"),
 dim_col = "o",
 origin_fn = centroid
# Specify dimming_fn
# around the centroid
dim_values(
 data = df,
 cols = c("x", "y"),
 dim_col = "o",
 origin_fn = centroid,
  dimming_fn = function(x, d) {
   x * 1 / (2^{(1 + d)})
  }
)
# Dim cluster-wise
# Group-wise dimming
df_dimmed <- df %>%
  dplyr::group_by(.cluster) %>%
  dim_values(
   cols = c("x", "y"),
   dim_col = "o",
    origin_fn = centroid
  )
# Plot the dimmed data such that the alpha (opacity) is
# controlled by the dimming
# (Note: This works because the `o` column is 1 for all values)
ggplot(
 data = df_dimmed,
 aes(x = x, y = y, alpha = o\_dimmed, color = .cluster)
```

30 distance

```
) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "Cluster", alpha = "o_dimmed")
```

distance

Calculate the distance to an origin

## Description

## [Experimental]

Calculates the distance to the specified origin with:

$$d(P1, P2) = sqrt((x2 - x1)^{2} + (y2 - y1)^{2} + (z2 - z1)^{2} + \dots)$$

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and finding the distance to e.g. the centroid of each group.

## Usage

```
distance(
  data,
  cols = NULL,
  origin = NULL,
  origin_fn = NULL,
  distance_col_name = ".distance",
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

## Arguments

overwrite

data	data.frame or vector.	
cols	Names of columns in `data` to measure distance in. Each column is considered a dimension.	
origin	Coordinates of the origin to calculate distances to. A scalar to use in all dimensions or a vector with one scalar per dimension.	
	<b>N.B.</b> Ignored when `origin_fn` is not NULL.	
origin_fn	Function for finding the origin coordinates.	
	<b>Input</b> : Each column will be passed as a vector in the order of `cols`.	
	Output: A vector with one scalar per dimension.	
	Can be created with <code>create_origin_fn()</code> if you want to apply the same function to each dimension.	
	E.g. `create_origin_fn(median)` would find the median of each column.	
	<pre>Built-in functions are centroid(), most_centered(), and midrange()</pre>	
distance_col_name		
	Name of new column with the distances.	
origin_col_name		

Whether to allow overwriting of existing columns. (Logical)

Name of new column with the origin coordinates. If NULL, no column is added.

#### Value

data.frame (tibble) with the additional columns (distances and origin coordinates).

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other measuring functions: angle(), vector_length()
Other distance functions: closest_to(), dim_values(), expand_distances_each(), expand_distances(), furthest_from(), swirl_2d(), swirl_3d()
```

#### **Examples**

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  x'' = runif(20),
  "y" = runif(20),
  "g" = rep(1:4, each = 5)
# Calculate distances in the two dimensions (x and y)
# With the origin at x=0.5, y=0.5
distance(
  data = df,
 cols = c("x", "y"),
  origin = c(0.5, 0.5)
# Calculate distances to the centroid for each group in 'g'
distance(
  data = dplyr::group_by(df, g),
 cols = c("x", "y"),
  origin_fn = centroid
```

expand\_distances

Expand the distances to an origin

#### **Description**

## [Experimental]

Moves the data points in n-dimensional space such that their distance to a specified origin is increased/decreased. A `multiplier` greater than 1 leads to expansion, while a positive `multiplier` lower than 1 leads to contraction.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data.frame and expanding around e.g. the centroid of each group.

The multiplier/exponent can be supplied as a constant or as a function that returns a constant. The latter can be useful when supplying a grouped data. frame and the multiplier/exponent depends on the data in the groups.

For expansion in each dimension separately, use expand\_distances\_each().

**NOTE**: When exponentiating, the default is to first add 1 to the distances, to ensure expansion even when the distance is between 0 and 1. If you need the purely exponentiated distances, disable `add\_one\_exp`.

#### Usage

```
expand_distances(
   data,
   cols = NULL,
   multiplier = NULL,
   multiplier_fn = NULL,
   origin = NULL,
   origin_fn = NULL,
   exponentiate = FALSE,
   add_one_exp = TRUE,
   suffix = "_expanded",
   keep_original = TRUE,
   mult_col_name = ifelse(isTRUE(exponentiate), ".exponent", ".multiplier"),
   origin_col_name = ".origin",
   overwrite = FALSE
)
```

## **Arguments**

data data.frame or vector.

cols Names of columns in `data` to expand coordinates of. Each column is consid-

ered a dimension.

multiplier Constant to multiply/exponentiate the distances to the origin by.

**N.B.** When 'exponentiate' is TRUE, the 'multiplier' becomes an *exponent*.

multiplier\_fn Function for finding the `multiplier`.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A numeric scalar.

origin Coordinates of the origin to expand around. A scalar to use in all dimensions or

a vector with one scalar per dimension.

**N.B.** Ignored when `origin\_fn` is not NULL.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

 $E.g. \ \verb|`create_origin_fn(median)| \ would find the median of each column.$ 

Built-in functions are centroid(), most\_centered(), and midrange()

exponentiate Whether to exponentiate instead of multiplying. (Logical)

add\_one\_exp Whether to add 1 to the distances before exponentiating to ensure they don't

contract when between  $\emptyset$  and 1. The added value is subtracted after the expo-

nentiation. (Logical)

The distances to the origin (`d`) are exponentiated as such:

d < -d + 1

d <- d ^ multiplier</pre>

d < -d - 1

 ${f N.B.}$  Ignored when `exponentiate` is FALSE.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

 $\verb| mult_col_name | Name of new column with the \verb|`multiplier"|. If \verb| NULL|, no column is added|.$ 

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### **Details**

Increases the distance to the origin in n-dimensional space by multiplying or exponentiating it by the multiplier.

We first move the origin to the zero-coordinates (e.g. c(0, 0, 0)) and normalize each vector to unit length. We then multiply this unit vector by the multiplied/exponentiated distance and moves the origin back to its original coordinates.

The distance to the specified origin is calculated with:

$$d(P1, P2) = sqrt((x2 - x1)^2 + (y2 - y1)^2 + (z2 - z1)^2 + \dots)$$

Note: By default (when `add\_one\_exp` is TRUE), we add 1 to the distance before the exponentiation and subtract it afterwards. See `add\_one\_exp`.

## Value

data.frame (tibble) with the expanded columns, along with the applied multiplier/exponent and origin coordinates.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()
```

Other expander functions: expand\_distances\_each()

```
Other distance functions: closest_to(), dim_values(), distance(), expand_distances_each(), furthest_from(), swirl_2d(), swirl_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(purrr)
library(ggplot2)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  x'' = runif(20),
  "y" = runif(20),
  "g" = rep(1:4, each = 5)
\# Expand distances in the two dimensions (x and y)
\# With the origin at x=0.5, y=0.5
# We multiply the distances by 2
expand_distances(
  data = df,
  cols = c("x", "y"),
 multiplier = 2,
 origin = c(0.5, 0.5)
\# Expand distances in the two dimensions (x and y)
# With the origin at x=0.5, y=0.5
# We exponentiate the distances by 2
expand_distances(
  data = df,
 cols = c("x", "y"),
 multiplier = 2,
 exponentiate = TRUE,
  origin = 0.5
)
# Expand values in one dimension (x)
\# With the origin at x=0.5
# We exponentiate the distances by 3
expand_distances(
  data = df,
  cols = c("x"),
  multiplier = 3,
  exponentiate = TRUE,
  origin = 0.5
\# Expand x and y around the centroid
# We use exponentiation for a more drastic effect
# The add_one_exp makes sure it expands
# even when x or y is in the range [0, <1]
# To compare multiple exponents, we wrap the
# call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
```

```
x = c(1, 3, 5),
  .f = function(exponent) {
    expand_distances(
      data = df,
      cols = c("x", "y"),
      multiplier = exponent,
      origin_fn = centroid,
      exponentiate = TRUE,
      add_one_exp = TRUE
   )
  }
)
df_expanded
\# Plot the expansions of x and y around the overall centroid
ggplot(df_expanded, aes(x = x_expanded, y = y_expanded, color = factor(.exponent))) +
  geom_vline(
    xintercept = df_expanded[[".origin"]][[1]][[1]],
    size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_hline(
   yintercept = df_expanded[[".origin"]][[1]][[2]],
    size = 0.2, alpha = .4, linetype = "dashed"
  geom_path(size = 0.2) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "Exponent")
# Expand x and y around the centroid using multiplication
# To compare multiple multipliers, we wrap the
# call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  x = c(1, 3, 5),
  .f = function(multiplier) {
    expand_distances(df,
      cols = c("x", "y"),
      multiplier = multiplier,
      origin_fn = centroid,
      exponentiate = FALSE
  }
)
df_expanded
\# Plot the expansions of x and y around the overall centroid
ggplot(df_expanded, aes(x = x_expanded, y = y_expanded, color = factor(.multiplier))) +
  geom_vline(
    xintercept = df_expanded[[".origin"]][[1]][[1]],
    size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_hline(
   yintercept = df_expanded[[".origin"]][[1]][[2]],
   size = 0.2, alpha = .4, linetype = "dashed"
  geom_path(size = 0.2, alpha = .8) +
  geom_point() +
```

```
theme_minimal() +
  labs(x = "x", y = "y", color = "Multiplier")
# Contraction
# Group-wise contraction to create clusters
df contracted <- df %>%
  dplyr::group_by(g) %>%
  expand_distances(
    cols = c("x", "y"),
    multiplier = 0.07,
    suffix = "_contracted",
    origin_fn = centroid
  )
# Plot the clustered data point on top of the original data points
ggplot(df\_contracted, aes(x = x\_contracted, y = y\_contracted, color = factor(g))) +
  geom_point(aes(x = x, y = y, color = factor(g)), alpha = 0.3, shape = 16) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "g")
```

expand\_distances\_each Expand the distances to an origin in each dimension

#### **Description**

#### [Experimental]

Moves the data points in n-dimensional space such that their distance to the specified origin is increased/decreased *in each dimension separately*. A `multiplier` greater than 1 leads to expansion, while a positive `multiplier` lower than 1 leads to contraction.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data.frame and expanding around e.g. the centroid of each group.

The multipliers/exponents can be supplied as constant(s) or as a function that returns constants. The latter can be useful when supplying a grouped data. frame and the multiplier/exponent depends on the data in the groups. If supplying multiple constants, there must be one per dimension (length of `cols`).

For expansion of the *multidimensional* distance, use expand\_distances().

**NOTE**: When exponentiating, the default is to first add 1 or -1 (depending on the sign of the distance) to the distances, to ensure expansion even when the distance is between -1 and 1. If you need the purely exponentiated distances, disable `add\_one\_exp`.

#### Usage

```
expand_distances_each(
  data,
  cols = NULL,
  multipliers = NULL,
```

```
multipliers_fn = NULL,
  origin = NULL,
  origin_fn = NULL,
  exponentiate = FALSE,
  add_one_exp = TRUE,
  suffix = "_expanded",
  keep_original = TRUE,
  mult_col_name = ifelse(isTRUE(exponentiate), ".exponents", ".multipliers"),
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

#### **Arguments**

data data.frame or vector.

cols Names of columns in `data` to expand. Each column is considered a dimension

to expand in.

use in all dimensions or a vector with one scalar per dimension.

**N.B.** When `exponentiate` is TRUE, the `multipliers` become *exponents*.

multipliers\_fn Function for finding the `multipliers`.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A numeric vector with one element per dimension.

Just as for `origin\_fn`, it can be created with create\_origin\_fn() if you want to apply the same function to each dimension. See `origin\_fn`.

origin Coordinates of the origin to expand around. A scalar to use in all dimensions or

a vector with one scalar per dimension.

**N.B.** Ignored when `origin\_fn` is not NULL.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

E.g. `create\_origin\_fn(median)` would find the median of each column.

Built-in functions are centroid(), most\_centered(), and midrange()

exponentiate Whether to exponentiate instead of multiplying. (Logical)

add\_one\_exp Whether to add the sign (either 1 or -1) before exponentiating to ensure the values don't contract. The added value is subtracted after the exponentiation.

(Logical)

Exponentiation becomes:

x < -x + sign(x)

x <- sign(x) \* abs(x) ^ multiplier</pre>

x <- x - sign(x)

**N.B.** Ignored when `exponentiate` is FALSE.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

 $\verb|mult_col_name| & Name of new column with the multiplier (s). If \verb|NULL|, no column is added. \\$ 

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### **Details**

For each value of each dimension (column), either multiply or exponentiate by the multiplier:

```
# Multiplication
x <- x * multiplier
# Exponentiation
x <- sign(x) * abs(x) ^ multiplier</pre>
```

Note: By default (when `add\_one\_exp` is TRUE), we add the sign (1 / -1) of the value before the exponentiation and subtract it afterwards. See `add\_one\_exp`.

#### Value

data.frame (tibble) with the expanded columns, along with the applied multiplier/exponent and origin coordinates.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()

Other expander functions: expand_distances()

Other distance functions: closest_to(), dim_values(), distance(), expand_distances(), furthest_from(), swirl_2d(), swirl_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(purrr)
library(ggplot2)

# Set seed
set.seed(1)

# Create a data frame
df <- data.frame(
   "x" = runif(20),
   "y" = runif(20),</pre>
```

```
"g" = rep(1:4, each = 5)
# Expand values in the two dimensions (x and y)
# With the origin at x=0.5, y=0.5
# We expand x by 2 and y by 4
expand_distances_each(
 data = df,
 cols = c("x", "y"),
 multipliers = c(2, 4),
 origin = c(0.5, 0.5)
# Expand values in the two dimensions (x and y)
# With the origin at x=0.5, y=0.5
# We expand both by 3
expand_distances_each(
  data = df,
  cols = c("x", "y"),
 multipliers = 3,
 origin = 0.5
# Expand values in one dimension (x)
# With the origin at x=0.5
# We expand by 3
expand_distances_each(
 data = df,
 cols = c("x"),
 multipliers = 3,
 origin = 0.5
# Expand x and y around the centroid
# We use exponentiation for a more drastic effect
# The add_one_exp makes sure it expands
# even when x or y is in the range [>-1, <1]
# To compare multiple exponents, we wrap the
# call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  x = c(1, 2.0, 3.0, 4.0),
  .f = function(exponent) {
    expand_distances_each(
      data = df,
      cols = c("x", "y"),
      multipliers = exponent,
      origin_fn = centroid,
      exponentiate = TRUE,
      add\_one\_exp = TRUE
   )
  }
df_expanded
\# Plot the expansions of x and y around the overall centroid
ggplot(df_expanded, aes(x = x_expanded, y = y_expanded, color = factor(.exponents))) +
  geom_vline(
```

```
xintercept = df_expanded[[".origin"]][[1]][[1]],
   size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_hline(
    yintercept = df_expanded[[".origin"]][[1]][[2]],
    size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "Exponent")
# Expand x and y around the centroid using multiplication
# To compare multiple multipliers, we wrap the
# call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  x = c(1, 2.0, 3.0, 4.0),
  .f = function(multiplier) {
    expand_distances_each(df,
      cols = c("x", "y"),
      multipliers = multiplier,
      origin_fn = centroid,
      exponentiate = FALSE
  }
)
df_expanded
\# Plot the expansions of x and y around the overall centroid
ggplot(df_expanded, aes(x = x_expanded, y = y_expanded, color = factor(.multipliers))) +
  geom vline(
    xintercept = df_expanded[[".origin"]][[1]][[1]],
    size = 0.2, alpha = .4, linetype = "dashed"
  ) +
    yintercept = df_expanded[[".origin"]][[1]][[2]],
    size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "Multiplier")
# Expand x and y with different multipliers
# around the centroid using multiplication
df_expanded <- expand_distances_each(</pre>
  df,
  cols = c("x", "y"),
  multipliers = c(1.25, 10),
 origin_fn = centroid,
  exponentiate = FALSE
)
df_expanded
# Plot the expansions of x and y around the overall centroid
# Note how the y axis is expanded a lot more than the x-axis
ggplot(df_expanded, aes(x = x_expanded, y = y_expanded)) +
    xintercept = df_expanded[[".origin"]][[1]][[1]],
```

FixedGroupsPipeline 41

```
size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_hline(
    yintercept = df_expanded[[".origin"]][[1]][[2]],
    size = 0.2, alpha = .4, linetype = "dashed"
  ) +
  geom_line(aes(color = "Expanded")) +
  geom_point(aes(color = "Expanded")) +
  geom\_line(aes(x = x, y = y, color = "Original")) +
  geom_point(aes(x = x, y = y, color = "Original")) +
  theme_minimal() +
  labs(x = "x", y = "y", color = "Multiplier")
#
 Contraction
#
# Group-wise contraction to create clusters
df contracted <- df %>%
  dplyr::group_by(g) %>%
  expand_distances_each(
    cols = c("x", "y"),
    multipliers = 0.07,
    suffix = "_contracted",
    origin_fn = centroid
# Plot the clustered data point on top of the original data points
ggplot(df\_contracted, aes(x = x\_contracted, y = y\_contracted, color = factor(g))) +
  geom_point(aes(x = x, y = y, color = factor(g)), alpha = 0.3, shape = 16) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "g")
```

FixedGroupsPipeline Chain multiple transformations with different argument values per group

### **Description**

### [Experimental]

Build a pipeline of transformations to be applied sequentially.

Specify different argument values for each group in a fixed set of groups. E.g. if your data.frame contains 5 groups, you provide 5 argument values for each of the non-constant arguments (see `var\_args`).

The number of expected groups is specified during initialization and the input `data` must be grouped such that it contains that exact number of groups.

Transformations are applied to groups separately, why the given transformation function only receives the subset of `data` belonging to the current group.

**Standard workflow**: Instantiate pipeline -> Add transformations -> Apply to data

To apply the same arguments to all groups, see Pipeline.

To apply generated argument values to an arbitrary number of groups, see GeneratedPipeline.

### Super class

```
rearrr::Pipeline -> FixedGroupsPipeline
```

#### **Public fields**

```
transformations \ \ list \ of \ transformations \ to \ apply.
```

names Names of the transformations.

num\_groups Number of groups the pipeline will be applied to.

#### Methods

### **Public methods:**

- FixedGroupsPipeline\$new()
- FixedGroupsPipeline\$add\_transformation()
- FixedGroupsPipeline\$apply()
- FixedGroupsPipeline\$print()
- FixedGroupsPipeline\$clone()

**Method** new(): Initialize the pipeline with the number of groups the pipeline will be applied to.

Usage:

```
FixedGroupsPipeline$new(num_groups)
```

Arguments:

num\_groups Number of groups the pipeline will be applied to.

 $\label{lem:method} \textbf{Method} \ \mbox{add\_transformation():} \ \ \mbox{Add a transformation to the pipeline}.$ 

Usage:

```
FixedGroupsPipeline$add_transformation(fn, args, var_args, name)
```

Arguments:

fn Function that performs the transformation.

args Named list with arguments for the `fn` function.

var\_args Named list of arguments with list of differing values for each group.

```
E.g. list("a" = list(1, 2, 3), "b" = list("a", "b", "c")) given 3 groups.
```

By adding ".apply" with a list of TRUE/FALSE flags, the transformation can be disabled for a specific group.

```
E.g. list(".apply" = list(TRUE, FALSE, TRUE), ....
```

name Name of the transformation step. Must be unique.

*Returns:* The pipeline. To allow chaining of methods.

**Method** apply(): Apply the pipeline to a data. frame.

Usage:

```
FixedGroupsPipeline$apply(data, verbose = FALSE)
```

Arguments:

data data. frame with the same number of groups as pre-registered in the pipeline.

You can find the number of groups in `data` with `dplyr::n\_groups(data)`. The number of groups expected by the pipeline can be accessed with `pipe\$num\_groups`.

verbose Whether to print the progress.

Returns: Transformed version of `data`.

FixedGroupsPipeline 43

```
Method print(): Print an overview of the pipeline.

Usage:
FixedGroupsPipeline$print(...)

Arguments:
... further arguments passed to or from other methods.

Returns: The pipeline. To allow chaining of methods.

Method clone(): The objects of this class are cloneable with this method.

Usage:
FixedGroupsPipeline$clone(deep = FALSE)

Arguments:
deep Whether to make a deep clone.
```

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

Other pipelines: GeneratedPipeline, Pipeline

```
# Attach package
library(rearrr)
library(dplyr)
# Create a data frame
# We group it by G so we have 3 groups
df <- data.frame(</pre>
  "Index" = 1:12,
  "A" = c(1:4, 9:12, 15:18),
  "G" = rep(1:3, each = 4)
) %>%
  dplyr::group_by(G)
# Create new pipeline
pipe <- FixedGroupsPipeline$new(num_groups = 3)</pre>
# Add 2D rotation transformation
pipe$add_transformation(
  fn = rotate_2d,
  args = list(
    x_col = "Index",
    y_col = "A",
    suffix = "",
    overwrite = TRUE
  ),
  var_args = list(
    degrees = list(45, 90, 180),
    origin = list(c(0, 0), c(1, 2), c(-1, 0))
  ),
  name = "Rotate"
```

```
)
# Add the `cluster_group` transformation
\# As the function is fed an ungrouped subset of `data`,
# i.e. the rows of that group, we need to specify `group_cols` in `args`
# That is specific to `cluster_groups()` though
# Also note `.apply` in `var_args` which tells the pipeline *not*
# to apply this transformation to the second group
pipe$add_transformation(
  fn = cluster_groups,
  args = list(
    cols = c("Index", "A"),
    suffix = "",
   overwrite = TRUE,
   group_cols = "G"
  ),
  var_args = list(
   multiplier = list(0.5, 1, 5),
    .apply = list(TRUE, FALSE, TRUE)
  ),
  name = "Cluster"
# Check pipeline object
# Apply pipeline to already grouped data.frame
# Enable `verbose` to print progress
pipe$apply(df, verbose = TRUE)
```

FixedGroupsTransformation

FixedGroupsTransformation

## **Description**

# [Experimental]

Container for the type of transformation used in FixedGroupsPipeline.

**Note**: For internal use.

#### Super class

```
rearrr::Transformation -> FixedGroupsTransformation
```

# **Public fields**

```
name Name of transformation.

fn Transformation function.

args list of constant arguments for `fn`.

var_args list of arguments for `fn` with different values per group.

num_groups Number of groups that the transformation expects.
```

apply\_arg list of TRUE/FALSE flags indicating whether the transformation should be applied to each of the groups.

When `NULL`, the transformation is applied to all groups.

### Methods

```
Public methods:
```

```
• FixedGroupsTransformation$new()
```

- FixedGroupsTransformation\$get\_group\_args()
- FixedGroupsTransformation\$apply()
- FixedGroupsTransformation\$print()
- FixedGroupsTransformation\$clone()

### **Method** new(): Initialize transformation.

```
Usage:
```

FixedGroupsTransformation\$new(fn, args, var\_args, name = NULL)

Arguments:

fn Transformation function.

args list of constant arguments for `fn`.

var\_args list of arguments for `fn` with different values per group. Each argument should have a list of values (one per group).

By adding ".apply" with a list of TRUE/FALSE flags, the transformation can be disabled for a specific group.

E.g. list(".apply" = list(TRUE, FALSE, TRUE), ....

name Name of transformation.

### **Method** get\_group\_args(): Get arguments for specific group ID.

Usage:

FixedGroupsTransformation\$get\_group\_args(group\_id)

Arguments:

group\_id ID of the group to get arguments for.

Returns: list of arguments.

## **Method** apply(): Apply the transformation to a data. frame.

Usage:

FixedGroupsTransformation\$apply(data)

Arguments:

data data. frame with the expected number of groups.

Returns: Transformed version of `data`.

### **Method** print(): Print an overview of the transformation.

Usage:

FixedGroupsTransformation\$print(..., indent = 0, show\_class = TRUE)

Arguments:

... further arguments passed to or from other methods.

indent How many spaces to indent when printing.

show\_class Whether to print the transformation class name.

46 flip\_values

Returns: The pipeline. To allow chaining of methods.

**Method** clone(): The objects of this class are cloneable with this method.

Usage:

FixedGroupsTransformation\$clone(deep = FALSE)

Arguments:

deep Whether to make a deep clone.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

Other transformation classes: GeneratedTransformation, Transformation

flip\_values

Flip the values around an origin

# Description

## [Experimental]

The values are flipped with the formula 'x = 2 \* c - x' where x is the value and c is the origin coordinate to flip the values around.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data.frame and flipping around e.g. the centroid of each group. By default the median value in each dimension is used.

The \*\_vec() version take and return a vector.

# **Example:**

```
The column values:
```

```
c(5, 2, 7, 4, 3, 1)
and the origin_fn = create_origin_fn(median)
Changes the values to:
c(2, 5, 0, 3, 4, 6)
```

### Usage

```
flip_values(
  data,
  cols = NULL,
  origin = NULL,
  origin_fn = create_origin_fn(median),
  suffix = "_flipped",
  keep_original = TRUE,
  origin_col_name = ".origin",
  overwrite = FALSE
)
flip_values_vec(data, origin = NULL, origin_fn = create_origin_fn(median))
```

flip\_values 47

### **Arguments**

data data.frame or vector. cols Names of columns in `data` to flip values of. origin Coordinates of the origin to flip around. A scalar to use in all dimensions (columns) or a vector with one scalar per dimension. **N.B.** Ignored when `origin\_fn` is not NULL. Remember to set it to NULL when passing origin coordinates manually! origin\_fn Function for finding the origin coordinates. **Input**: Each column will be passed as a vector in the order of `cols`. Output: A vector with one scalar per dimension. Can be created with create\_origin\_fn() if you want to apply the same function to each dimension. E.g. `create\_origin\_fn(median)` would find the median of each column. **Built-in functions** are centroid(), most\_centered(), and midrange() suffix Suffix to add to the names of the generated columns. Use an empty string (i.e. "") to overwrite the original columns. keep\_original Whether to keep the original columns. (Logical) Some columns may have been overwritten, in which case only the newest versions are returned. origin\_col\_name Name of new column with the origin coordinates. If NULL, no column is added.

### Author(s)

overwrite

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()
```

Whether to allow overwriting of existing columns. (Logical)

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)

# Set seed
set.seed(1)

# Create a data frame
df <- data.frame(
   "Index" = 1:10,
   "A" = sample(1:10),
   "B" = runif(10),
   "G" = c(
    1, 1, 1, 2, 2,
    2, 3, 3, 3, 3</pre>
```

48 furthest\_from

```
),
  stringsAsFactors = FALSE
# Flip values of the columns
flip_values(df$A)
flip_values(df, cols = "A")
flip_values(df, cols = "B", origin = 0.3, origin_fn = NULL, keep_original = FALSE)
flip_values(df,
  cols = c("A", "B"),
  origin = c(3, 0.3),
  origin_fn = NULL,
  suffix = "",
 keep_original = FALSE,
  overwrite = TRUE
flip_values(df, cols = c("A", "B"), origin_fn = create_origin_fn(max))
# Grouped by G
df %>%
  dplyr::group_by(G) %>%
  flip_values(
    cols = c("A", "B"),
    origin_fn = create_origin_fn(median),
    keep_original = FALSE
# Plot A and flipped A
# First flip A around the median and then around the value 3.
df <- df %>%
  flip_values(cols = "A", suffix = "_flip_median", origin_col_name = NULL) %>%
  flip_values(cols = "A", suffix = "_flip_3", origin = 3,
              origin_fn = NULL, origin_col_name = NULL)
# Plot A and A flipped around its median
ggplot(df, aes(x = Index, y = A)) +
  geom_line(aes(color = "A")) +
  geom_line(aes(y = A_flip_median, color = "Flipped A (median)")) +
  geom_hline(aes(color = "Median A", yintercept = median(A))) +
  theme_minimal()
# Plot A and A flipped around the value 3
ggplot(df, aes(x = Index, y = A)) +
  geom_line(aes(color = "A")) +
  geom_line(aes(y = A_flip_3, color = "Flipped A (3)")) +
  geom_hline(aes(color = "3", yintercept = 3)) +
  theme_minimal()
```

furthest\_from

Orders values by longest distance to an origin

### **Description**

[Experimental]

furthest\_from 49

Values are ordered by how far they are from the origin.

In 1d (when `cols` has length 1), the origin can be thought of as a target value. In n dimensions, the origin can be thought of as coordinates.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data.frame and ordering the rows by their distance to the centroid of each group.

The \*\_vec() version takes and returns a vector.

### **Example:**

```
The column values:
```

```
c(1, 2, 3, 4, 5)
and origin = 2
```

### are ordered as:

```
c(5, 4, 1, 3, 2)
```

# Usage

```
furthest_from(
  data,
  cols = NULL,
  origin = NULL,
  origin_fn = NULL,
  shuffle_ties = FALSE,
  origin_col_name = ".origin",
  distance_col_name = ".distance",
  overwrite = FALSE
)
```

furthest\_from\_vec(data, origin = NULL, origin\_fn = NULL, shuffle\_ties = FALSE)

### **Arguments**

data data.frame or vector.

cols Column(s) to create sorting factor by. When `NULL` and `data` is a data. frame,

the row numbers are used.

origin Coordinates of the origin to calculate distances to. A scalar to use in all dimen-

sions or a vector with one scalar per dimension.

**N.B.** Ignored when `origin\_fn` is not `NULL`.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same func-

tion to each dimension.

 $E.g. \ \verb|`create_origin_fn(median)| \ would find the median of each column.$ 

**Built-in functions** are centroid(), most\_centered(), and midrange()

shuffle\_ties
origin\_col\_name

Whether to shuffle elements with the same distance to the origin. (Logical)

Name of new column with the origin coordinates. If `NULL`, no column is added.

50 furthest\_from

```
distance_col_name
```

Name of new column with the distances to the origin. If `NULL`, no column is added.

overwrite

Whether to allow overwriting of existing columns. (Logical)

### Value

The sorted data.frame (tibble) / vector.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), pair_extremes(), position_max(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

Other distance functions: closest\_to(), dim\_values(), distance(), expand\_distances\_each(), expand\_distances(), swirl\_2d(), swirl\_3d()

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "index" = 1:10,
  "A" = sample(1:10),
  "B" = runif(10),
  G'' = c(
   1, 1, 1, 2, 2,
   2, 3, 3, 3, 3
  ),
  stringsAsFactors = FALSE
# Furthest from 3 in a vector
furthest_from_vec(1:10, origin = 3)
# Furthest from the third row (index of data.frame)
furthest_from(df, origin = 3)$index
# By each of the columns
furthest_from(df, cols = "A", origin = 3)$A
furthest_from(df, cols = "A", origin_fn = most_centered)$A
furthest_from(df, cols = "B", origin = 0.5)$B
furthest_from(df, cols = "B", origin_fn = centroid)$B
# Shuffle the elements with the same distance to the origin
furthest_from(df,
```

GeneratedPipeline 51

```
cols = "A",
  origin_fn = create_origin_fn(median),
  shuffle_ties = TRUE
) $A
# Grouped by G
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  furthest_from(
   cols = "A",
    origin_fn = create_origin_fn(median)
  )
# Plot the rearranged values
plot(
  x = 1:10,
  y = furthest_from(df,
   cols = "B",
   origin_fn = create_origin_fn(median)
  xlab = "Position", ylab = "B"
plot(
  x = 1:10,
  y = furthest_from(df,
   cols = "A",
    origin_fn = create_origin_fn(median),
    shuffle_ties = TRUE
  )$A,
  xlab = "Position", ylab = "A"
# In multiple dimensions
df %>%
  furthest_from(cols = c("A", "B"), origin_fn = most_centered)
```

GeneratedPipeline

Chain multiple transformations and generate argument values per group

### **Description**

### [Experimental]

Build a pipeline of transformations to be applied sequentially.

Generate argument values for selected arguments with a given set of generators. E.g. randomly generate argument values for each group in a data. frame.

Groupings are reset between each transformation. See group\_cols.

**Standard workflow**: Instantiate pipeline -> Add transformations -> Apply to data

To apply the same arguments to all groups, see Pipeline.

To apply different but specified argument values to a fixed set of groups, see FixedGroupsPipeline.

52 GeneratedPipeline

### Super class

```
rearrr::Pipeline -> GeneratedPipeline
```

#### **Public fields**

transformations list of transformations to apply. names Names of the transformations.

#### Methods

#### **Public methods:**

- GeneratedPipeline\$add\_transformation()
- GeneratedPipeline\$print()
- GeneratedPipeline\$clone()

Method add\_transformation(): Add a transformation to the pipeline.

```
Usage:
```

```
GeneratedPipeline$add_transformation(
   fn,
   args,
   generators,
   name,
   group_cols = NULL
)
```

Arguments:

fn Function that performs the transformation.

args Named list with arguments for the `fn` function.

generators Named list of functions for generating argument values for a single call of `fn`. It is possible to include an *apply generator* for deciding whether the transformation should be applied to the current group or not. This is done by adding a function with the name `.apply` to the `generators` list. E.g. ".apply" = function(){sample(c(TRUE, FALSE), 1)}.

name Name of the transformation step. Must be unique.

group\_cols Names of the columns to group the input data by before applying the transformation

Note that the transformation function is applied separately to each group (subset). If the 'fn' function requires access to the entire data.frame, the grouping columns should be specified as part of 'args' and handled by the 'fn' function.

Returns: The pipeline. To allow chaining of methods.

## Examples:

Method print(): Print an overview of the pipeline.

Usage:

GeneratedPipeline 53

```
GeneratedPipeline$print(...)

Arguments:
... further arguments passed to or from other methods.

Returns: The pipeline. To allow chaining of methods.

Method clone(): The objects of this class are cloneable with this method.

Usage:
GeneratedPipeline$clone(deep = FALSE)

Arguments:
deep Whether to make a deep clone.
```

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

Other pipelines: FixedGroupsPipeline, Pipeline

```
# Attach package
library(rearrr)
# Create a data frame
df <- data.frame(</pre>
  "Index" = 1:12,
  "A" = c(1:4, 9:12, 15:18),
  "G" = rep(1:3, each = 4)
# Create new pipeline
pipe <- GeneratedPipeline$new()</pre>
# Add 2D rotation transformation
# Note that we specify the grouping via `group_cols`
pipe$add_transformation(
  fn = rotate_2d,
  args = list(
   x_{col} = "Index",
   y_{col} = "A",
    suffix = "",
    overwrite = TRUE
  generators = list(degrees = function(){sample.int(360, 1)},
                    origin = function(){rnorm(2)}),
  name = "Rotate",
  group_cols = "G"
# Add the `cluster_group` transformation
# Note that this function requires the entire input data
# to properly scale the groups. We therefore specify `group_cols`
# as part of `args`. This works as `cluster_groups()` accepts that
```

54 GeneratedTransformation

```
# argument.
\# Also note the `.apply` generator which generates a TRUE/FALSE scalar
# for whether the transformation should be applied to the current group
pipe$add_transformation(
  fn = cluster_groups,
  args = list(
    cols = c("Index", "A"),
    suffix = "",
    overwrite = TRUE.
   group_cols = "G"
  ),
  generators = list(
    multiplier = function() {
     0.1 * runif(1) * 3 ^ sample.int(5, 1)
   },
    .apply = function(){sample(c(TRUE, FALSE), 1)}
  ),
  name = "Cluster"
# Check pipeline object
pipe
# Apply pipeline to data.frame
# Enable `verbose` to print progress
pipe$apply(df, verbose = TRUE)
## Method `GeneratedPipeline$add_transformation`
# `generators` is a list of functions for generating
# argument values for a chosen set of arguments
# `.apply` can be used to disable the transformation
generators = list(degrees = function(){sample.int(360, 1)},
                  origin = function(){rnorm(2)},
                  .apply = function(){sample(c(TRUE, FALSE), 1)})
```

GeneratedTransformation

Generated Transformation

## **Description**

# [Experimental]

Container for the type of transformation used in GeneratedPipeline.

Note: For internal use.

## Super class

rearrr::Transformation -> GeneratedTransformation

**GeneratedTransformation** 55

#### **Public fields**

```
name Name of transformation.
fn Transformation function.
args list of constant arguments for `fn`.
generators list of generator functions for generating argument values.
apply_generator Generator function for deciding whether to apply the transformation to the cur-
     rent group.
```

### Methods

#### **Public methods:**

```
• GeneratedTransformation$new()
```

- GeneratedTransformation\$get\_group\_args()
- GeneratedTransformation\$generate\_args()
- GeneratedTransformation\$print()
- GeneratedTransformation\$clone()

Does not include the constant arguments.

```
Method new(): Initialize transformation.
 GeneratedTransformation$new(
   fn,
   args,
   generators,
   name = NULL,
   group_cols = NULL
 )
 Arguments:
 fn Transformation function.
 args list of constant arguments for `fn`.
 generators Named list of functions for generating argument values for a single call of `fn`.
     It is possible to include an apply generator for deciding whether the transformation should
     be applied to the current group or not. This is done by adding a function with the name
     `.apply` to the `generators` list. E.g. ".apply" = function(){sample(c(TRUE, FALSE),
     1)}.
 name Name of transformation.
 group_cols Names of columns to group data.frame by before applying `fn`.
     When `NULL`, the data. frame is not grouped.
Method get_group_args(): Get arguments for a group.
 Usage:
 GeneratedTransformation$get_group_args()
 Returns: list of arguments (both constant and generated).
Method generate_args(): Generate arguments for a group with the `generators`.
 Usage:
 GeneratedTransformation$generate_args()
 Returns: list of generated arguments.
```

56 generate\_clusters

```
Method print(): Print an overview of the transformation.

Usage:
GeneratedTransformation$print(..., indent = 0, show_class = TRUE)

Arguments:
... further arguments passed to or from other methods.
indent How many spaces to indent when printing.
show_class Whether to print the transformation class name.

Returns: The pipeline. To allow chaining of methods.

Method clone(): The objects of this class are cloneable with this method.

Usage:
GeneratedTransformation$clone(deep = FALSE)

Arguments:
deep Whether to make a deep clone.
```

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

 $Other \ transformation \ classes: \ Fixed Groups Transformation, \ Transformation$ 

generate\_clusters

Generate n-dimensional clusters

# **Description**

## [Experimental]

Generates data.frame (tibble) with clustered groups.

## Usage

```
generate_clusters(
  num_rows,
  num_cols,
  num_clusters,
  compactness = 1.6,
  generator = runif,
  name_prefix = "D",
  cluster_col_name = ".cluster"
)
```

generate\_clusters 57

### **Arguments**

Number of rows. num\_rows num\_cols Number of columns (dimensions). num\_clusters Number of clusters. How compact the clusters should be. A larger value leads to more compact compactness clusters (on average). Technically, it is passed to the `multiplier` argument in cluster\_groups() as '0.1/compactness'. Function to generate the numeric values. generator Must have the *number of values to generate* as its first (and only required) argument, as that is the only argument we pass to it. name\_prefix Prefix string for naming columns.

cluster\_col\_name

Name of cluster factor.

### **Details**

- Generates data. frame with random values using the `generator`.
- Divides the rows into groups (the clusters).
- Contracts the distance from each data point to the centroid of its group.
- Performs MinMax scaling such that the scale of the data points is similar to the generated data.

### Value

data.frame (tibble) with the clustered columns and the cluster grouping factor.

### Author(s)

```
Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>
```

## See Also

```
Other clustering functions: cluster_groups(), transfer_centroids()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
library(plotly)

# Set seed
set.seed(10)

# Generate clusters
generate_clusters(num_rows = 20, num_cols = 3, num_clusters = 3, compactness = 1.6)
generate_clusters(num_rows = 20, num_cols = 5, num_clusters = 6, compactness = 2.5)

# Generate clusters and plot them
# Tip: Call this multiple times
```

58 hexagonalize

```
# to see the behavior of `generate_clusters()`
generate_clusters(
  num_rows = 50, num_cols = 2,
  num_clusters = 5, compactness = 1.6
) %>%
  ggplot(
   aes(x = D1, y = D2, color = .cluster)
  ) +
  geom_point() +
  theme_minimal() +
  labs(x = "D1", y = "D2", color = "Cluster")
# Plot clusters in 3d view
#
# Generate clusters
clusters <- generate_clusters(</pre>
 num_rows = 50, num_cols = 3,
 num_clusters = 5, compactness = 1.6
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
 x = clusters$D1,
 y = clusters$D2,
 z = clusters$D3,
 type = "scatter3d",
 mode = "markers",
 color = clusters$.cluster
## End(Not run)
```

hexagonalize

Create x-coordinates so the points form a hexagon

# Description

### [Experimental]

Create the x-coordinates for a vector of y-coordinates such that they form a hexagon.

This will likely look most like a hexagon when the y-coordinates are somewhat equally distributed, e.g. a uniform distribution.

### Usage

```
hexagonalize(
  data,
  y_col = NULL,
  .min = NULL,
  .max = NULL,
  offset_x = 0,
```

hexagonalize 59

```
keep_original = TRUE,
  x_col_name = ".hexagon_x",
  edge_col_name = ".edge",
  overwrite = FALSE
)
```

# **Arguments**

data	data.frame or vector.
y_col	Name of column in `data` with y-coordinates to create x-coordinates for.
.min	Minimum y-coordinate. If NULL, it is inferred by the given y-coordinates.
.max	Maximum y-coordinate. If NULL, it is inferred by the given y-coordinates.
offset_x	Value to offset the x-coordinates by.
keep_original	Whether to keep the original columns. (Logical)
	Some columns may have been overwritten, in which case only the newest versions are returned.
x_col_name	Name of new column with the x-coordinates.
edge_col_name	Name of new column with the edge identifiers. If NULL, no column is added.
	Numbering is clockwise and starts at the upper-right edge.
overwrite	Whether to allow overwriting of existing columns. (Logical)

## Value

data.frame (tibble) with the added x-coordinates and an identifier for the edge the data point is a part of.

# Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

# See Also

Other forming functions: circularize(), square(), triangularize()

```
# Attach packages
library(rearrr)
library(dplyr)
library(gprose)

# Set seed
set.seed(1)

# Create a data frame
df <- data.frame(
   "y" = runif(200),
   "g" = factor(rep(1:5, each = 40))
)

# Hexagonalize 'y'</pre>
```

60 hexagonalize

```
df_hex <- hexagonalize(df, y_col = "y")</pre>
df_hex
# Plot hexagon
df_hex %>%
  ggplot(aes(x = .hexagon_x, y = y, color = .edge)) +
  geom_point() +
 theme_minimal()
# Grouped hexagonalization
# Hexagonalize 'y' for each group
# First cluster the groups a bit to move the
# hexagons away from each other
df_hex <- df %>%
 cluster_groups(
   cols = "y",
   group_cols = "g",
    suffix = "",
   overwrite = TRUE
  ) %>%
  dplyr::group_by(g) %>%
  hexagonalize(
   y_{col} = "y",
   overwrite = TRUE
  )
# Plot hexagons
df_hex %>%
  ggplot(aes(x = .hexagon_x, y = y, color = g)) +
  geom_point() +
  theme_minimal()
# Specifying minimum value
# Specify minimum value manually
df_hex <- hexagonalize(df, y_col = "y", .min = -2)
df_hex
# Plot hexagon
df_hex %>%
  ggplot(aes(x = .hexagon_x, y = y, color = .edge)) +
  geom_point() +
  theme_minimal()
# Multiple hexagons by contraction
# Start by hexagonalizing 'y'
df_hex <- hexagonalize(df, y_col = "y")</pre>
# Contract '.hexagon_x' and 'y' towards the centroid
```

is\_most\_centered 61

```
# To contract with multiple multipliers at once,
# we wrap the call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  x = c(1, 0.75, 0.5, 0.25, 0.125),
  .f = function(mult) {
    expand_distances(
      data = df_hex,
      cols = c(".hexagon_x", "y"),
      multiplier = mult,
      origin_fn = centroid,
      overwrite = TRUE
  }
df_expanded
df_expanded %>%
  ggplot(aes(
   x = .hexagon_x_expanded, y = y_expanded,
    color = .edge, alpha = .multiplier
  geom_point() +
  theme_minimal()
```

 $is\_most\_centered$ 

Find which data point is closest to the centroid

# Description

### [Experimental]

Finds the data point with the shortest distance to the centroid.

To get the coordinates of the most centered data point, use most\_centered() instead.

## Usage

```
is_most_centered(..., na.rm = FALSE)
```

## **Arguments**

... Numeric vectors.
na.rm Whether to ignore mis

Whether to ignore missing values. At least one data point must be complete. (Logical)

## Value

Logical vector (TRUE/FALSE) indicating if a data point is the most centered.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

62 median\_index

#### See Also

Other coordinate functions: centroid(), create\_origin\_fn(), midrange(), most\_centered()

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create three vectors
x \leftarrow runif(10)
y <- runif(10)</pre>
z \leftarrow runif(10)
# Find the data point closest to the centroid
is_most_centered(x, y, z)
# Compare to coordinates for the most centered
most_centered(x, y, z)
#
# For data.frames
# Create data frame
df <- data.frame(</pre>
  x'' = x
  "y" = y,
  z'' = z
  "g" = rep(1:2, each = 5)
# Filter the data points
# closest to the centroid
df %>%
 dplyr::filter(is_most_centered(x, y, z))
# When 'df' is grouped
df %>%
  dplyr::group_by(g) %>%
  dplyr::filter(is_most_centered(x, y, z))
# Add as column
df %>%
  dplyr::group_by(g) %>%
  dplyr::mutate(mc = is_most_centered(x, y, z))
```

median\_index 63

## **Description**

# [Experimental]

Applies function to the indices of each vector in `...`.

These functions were created with create\_n\_fn().

## Usage

```
median_index(..., negate = FALSE, round_fn = round)
quantile_index(..., prob, type = 7, negate = FALSE, round_fn = round)
```

## **Arguments**

... Numeric vectors.

negate Whether to negate the result. I.e. to multiply it with -1. (Logical) round\_fn Function for rounding output. Rounding is done *prior* to negation.

E.g. round, floor, or ceiling.

prob Probability in [0,1] for quantile().
type Quantile algorithm to use. See quantile().

### Value

numeric vector with one element per supplied vector.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

```
Other n functions: create_n_fn()
```

```
# Attach packages
library(rearrr)

# Set seed
set.seed(1)

# Create three vectors
x <- runif(10)
y <- runif(15)
z <- runif(20)

median_index(x, y, z)
quantile_index(x, y, z, prob = 0.2)

# Negate result
median_index(x, y, z, negate = TRUE)</pre>
```

64 midrange

midrange

Find the midrange values

## **Description**

# [Experimental]

Calculates the midrange for each of the passed vectors/columns.

Midrange is defined as:

```
(maxx + minx)/2
```

# Usage

```
midrange(..., cols = NULL, na.rm = FALSE)
```

### **Arguments**

```
    ... Numeric vectors or a single data.frame.
    cols Names of columns to use when `...` is a single data.frame.
    na.rm Whether to ignore missing values when calculating min and max values. (Logical)
```

### Value

Either a vector with the midrange of each supplied vector or a data. frame with the midrange of each supplied column along with any grouping variables.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

 $Other coordinate functions: centroid(), create\_origin\_fn(), is\_most\_centered(), most\_centered(), most\_cent$ 

```
# Attach packages
library(rearrr)
library(dplyr)

# Set seed
set.seed(1)

# Create three vectors
x <- runif(10)
y <- runif(10)
z <- runif(10)

# Find midrange for each vector
midrange(x, y, z)</pre>
```

min\_max\_scale 65

```
# For data.frames
#
# Create data frame
df <- data.frame(
    "x" = x,
    "y" = y,
    "z" = z,
    "g" = rep(1:2, each = 5)
)
# Find midrange for each column
midrange(df, cols = c("x", "y", "z"))
# When 'df' is grouped
df %>%
    dplyr::group_by(g) %>%
    midrange(cols = c("x", "y", "z"))
```

min\_max\_scale

Scale to a range

# Description

# [Experimental]

Scales the values to a range with MinMax scaling.

# Usage

```
min_max_scale(
    x,
    new_min,
    new_max,
    old_min = NULL,
    old_max = NULL,
    na.rm = FALSE
)
```

# Arguments

X	Numeric vector to scale.
new_min	Minimum value of target range.
new_max	Maximum value of target range.
old_min	Minimum value of original range.
	If NULL, this is the minimum value in `x`.
old_max	Maximum value of original range.
	If NULL, this is the maximum value in `x`.
na.rm	Whether missing values should be removed when calculating `old_min` and/or `old_max`.
	<b>N.B.</b> Ignored when both `old_min` and `old_max` are NULL.

66 most\_centered

#### Value

Scaled version of `x`.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

Other scaling functions: to\_unit\_length()

# **Examples**

```
# Attach packages
library(rearrr)

# Set seed
set.seed(1)

# Create numeric vector
x <- runif(10)

# Scale
min_max_scale(x, new_min = -1, new_max = 0)
min_max_scale(x, new_min = -1, new_max = 0, old_max = 3)</pre>
```

most\_centered

Find the coordinates for the data point closest to the centroid

## Description

# [Experimental]

Returns the coordinates for the data point with the shortest distance to the centroid.

To get a logical vector (TRUE/FALSE) indicating whether a data point is the most centered, use is\_most\_centered().

### Usage

```
most_centered(..., cols = NULL, na.rm = FALSE)
```

## **Arguments**

... Numeric vectors or a single data.frame.
 cols Names of columns to use when `...` is a single data.frame.
 na.rm Whether to ignore missing values. At least one data point must be complete. (Logical)

## Value

The coordinates for the data point closest to the centroid. Either as a vector or a data. frame.

most\_centered 67

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

Other coordinate functions: centroid(), create\_origin\_fn(), is\_most\_centered(), midrange()

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create three vectors
x \leftarrow runif(10)
y <- runif(10)
z <- runif(10)</pre>
# Find coordinates of the data point
# closest to the centroid
most_centered(x, y, z)
# Compare to centroid coordinates
centroid(x, y, z)
# For data.frames
# Create data frame
df <- data.frame(</pre>
  x'' = x
  y'' = y
  z'' = z
  "g" = rep(1:2, each = 5)
# Find coordinates of the data point
# closest to the centroid
most\_centered(df, cols = c("x", "y", "z"))
# When 'df' is grouped
df %>%
  dplyr::group_by(g) %>%
  most_centered(cols = c("x", "y", "z"))
# Filter to only include most centered data points
df %>%
  dplyr::group_by(g) %>%
  dplyr::filter(is_most_centered(x, y, z))
```

68 pair\_extremes

pair\_extremes

Pair extreme values and sort by the pairs

## **Description**

### [Experimental]

The values are paired/grouped such that the lowest and highest values form the first group, the second lowest and the second highest values form the second group, and so on. The values are then sorted by these groups/pairs.

When `data` has an uneven number of rows, the `unequal\_method` determines which group should have only 1 element.

The \*\_vec() version takes and returns a vector.

### Example:

```
The column values:
```

```
c(1, 2, 3, 4, 5, 6)
```

Creates the **sorting factor**:

```
c(1, 2, 3, 3, 2, 1)
```

And are ordered as:

```
c(1, 6, 2, 5, 3, 4)
```

## Usage

```
pair_extremes(
  data,
  col = NULL,
  unequal_method = "middle",
  num_pairings = 1,
  balance = "mean",
  order_by_aggregates = FALSE,
  shuffle_members = FALSE,
  shuffle_pairs = FALSE,
  factor_name = ifelse(num_pairings == 1, ".pair", ".pairing"),
  overwrite = FALSE
)
pair_extremes_vec(
  unequal_method = "middle",
  num_pairings = 1,
  balance = "mean",
  order_by_aggregates = FALSE,
  shuffle_members = FALSE,
  shuffle_pairs = FALSE
```

pair\_extremes 69

### **Arguments**

data data.frame or vector.

Column to create sorting factor by. When `NULL` and `data` is a data. frame, col

the row numbers are used.

unequal\_method Method for dealing with an unequal number of rows/elements in `data`.

One of: first, middle or last

**first:** The first group will have size 1.

### **Example:**

The ordered column values:

c(1, 2, 3, 4, 5)

Creates the **sorting factor**:

c(1, 2, 3, 3, 2)

And are **ordered as**:

c(1, 2, 5, 3, 4)

middle: The middle group will have size 1.

#### Example:

The ordered column values:

c(1, 2, 3, 4, 5)

Creates the **sorting factor**:

c(1, 3, 2, 3, 1)

And are **ordered as**:

c(1, 5, 3, 2, 4)

**last:** The last group will have size 1.

#### Example:

The ordered column values:

c(1, 2, 3, 4, 5)

Creates the **sorting factor**:

c(1, 2, 2, 1, 3)

And are **ordered as**:

c(1, 4, 2, 3, 5)

Number of pairings to perform (recursively). At least 1. num\_pairings

Based on 'balance', the secondary pairings perform extreme pairing on either

What to balance pairs for in a given secondary pairing. Either "mean", "spread",

the sum, absolute difference, min, or max of the pair elements.

"min", or "max". Can be a single string used for all secondary pairings or one

for each secondary pairing (`num\_pairings` - 1).

The first pairing always pairs the actual element values.

mean: Pairs have similar means. The values in the pairs from the previous pairing are aggregated with `sum()` and paired.

**spread:** Pairs have similar spread (e.g. standard deviations). The values in the pairs from the previous pairing are aggregated with `sum(abs(diff()))` and paired.

min / max: Pairs have similar minimum / maximum values. The values in the pairs from the previous pairing are aggregated with `min()` / `max()` and paired.

### order\_by\_aggregates

balance

Whether to order the pairs from initial pairings (first `num\_pairings` - 1) by their aggregate values instead of their pair identifiers.

N.B. Only used when `num\_pairings` > 1.

70 pair\_extremes

```
shuffle_members
```

Whether to shuffle the order of the group members within the groups. (Logical)

shuffle\_pairs Whether to shuffle the order of the pairs. Pair members remain together. (Logi-

cal)

factor\_name Name of new column with the sorting factor. If `NULL`, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

### Value

The sorted data. frame (tibble) / vector. Optionally with the sorting factor added.

When `data` is a vector and `keep\_factors` is FALSE, the output will be a vector. Otherwise, a data. frame.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), position_max(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "index" = 1:10,
  "A" = sample(1:10),
  "B" = runif(10),
  "C" = LETTERS[1:10],
  G'' = c(
   1, 1, 1, 2, 2,
   2, 3, 3, 3, 3
  ),
  stringsAsFactors = FALSE
# Pair extreme indices (row numbers)
pair_extremes(df)
# Pair extremes in each of the columns
pair_extremes(df, col = "A")$A
pair_extremes(df, col = "B")$B
pair_extremes(df, col = "C")$C
# Shuffle the members pair-wise
# The rows within each pair are shuffled
```

Pipeline 71

```
# while the `.pair` column maintains it order
pair_extremes(df, col = "A", shuffle_members = TRUE)
# Shuffle the order of the pairs
# The rows within each pair maintain their order
# and stay together but the `.pair` column is shuffled
pair_extremes(df, col = "A", shuffle_pairs = TRUE)
# Use recursive pairing
# Mostly meaningful with much larger datasets
# Order initial grouping by pair identifiers
pair_extremes(df, col = "A", num_pairings = 2)
# Order initial grouping by aggregate values
pair_extremes(df, col = "A", num_pairings = 2, order_by_aggregates = TRUE)
# Grouped by G
# Each G group only has 3 elements
# so it only creates 1 pair and a group
# with the single excessive element
# per G group
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  pair_extremes(col = "A")
# Plot the extreme pairs
plot(
 x = 1:10,
 y = pair_extremes(df, col = "B")$B,
 col = as.character(rep(1:5, each = 2))
# With shuffled pair members (run a few times)
plot(
  x = 1:10,
 y = pair_extremes(df, col = "B", shuffle_members = TRUE)$B,
  col = as.character(rep(1:5, each = 2))
# With shuffled pairs (run a few times)
plot(
 x = rep(1:5, each = 2),
 y = pair_extremes(df, col = "B", shuffle_pairs = TRUE)$B,
 col = as.character(rep(1:5, each = 2))
```

Pipeline

Chain multiple transformations

### **Description**

### [Experimental]

Build a pipeline of transformations to be applied sequentially.

Uses the same arguments for all groups in `data`.

Groupings are reset between each transformation. See group\_cols.

72 Pipeline

Standard workflow: Instantiate pipeline -> Add transformations -> Apply to data

To apply different argument values to each group, see GeneratedPipeline for generating argument values for an arbitrary number of groups and FixedGroupsPipeline for specifying specific values for a fixed set of groups.

#### **Public fields**

transformations list of transformations to apply. names Names of the transformations.

#### Methods

#### **Public methods:**

- Pipeline\$add\_transformation()
- Pipeline\$apply()
- Pipeline\$print()
- Pipeline\$clone()

**Method** add\_transformation(): Add a transformation to the pipeline.

```
Usage:
```

Pipeline\$add\_transformation(fn, args, name, group\_cols = NULL)

Arguments.

fn Function that performs the transformation.

args Named list with arguments for the `fn` function.

name Name of the transformation step. Must be unique.

group\_cols Names of the columns to group the input data by before applying the transformation.

Note that the transformation function is applied separately to each group (subset). If the 'fn' function requires access to the entire data.frame, the grouping columns should be specified as part of 'args' and handled by the 'fn' function.

Returns: The pipeline. To allow chaining of methods.

Method apply(): Apply the pipeline to a data. frame.

```
Usage:
```

Pipeline\$apply(data, verbose = FALSE)

Arguments:

data data.frame.

A grouped data.frame will raise a warning and the grouping will be ignored. Use the `group\_cols` argument in the `add\_transformation` method to specify how `data` should be grouped for each transformation.

verbose Whether to print the progress.

Returns: Transformed version of `data`.

Method print(): Print an overview of the pipeline.

```
Usage:
```

```
Pipeline$print(...)
```

Arguments:

... further arguments passed to or from other methods.

Pipeline 73

Returns: The pipeline. To allow chaining of methods.

**Method** clone(): The objects of this class are cloneable with this method.

```
Usage:
Pipeline$clone(deep = FALSE)
Arguments:
```

deep Whether to make a deep clone.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

Other pipelines: FixedGroupsPipeline, GeneratedPipeline

```
# Attach package
library(rearrr)
# Create a data frame
df <- data.frame(</pre>
  "Index" = 1:12,
  "A" = c(1:4, 9:12, 15:18),
  "G" = rep(1:3, each = 4)
# Create new pipeline
pipe <- Pipeline$new()</pre>
# Add 2D rotation transformation
# Note that we specify the grouping via `group_cols`
pipe$add_transformation(
  fn = rotate_2d,
  args = list(
   x_{col} = "Index",
   y_{col} = "A",
   origin = c(0, 0),
   degrees = 45,
   suffix = "",
   overwrite = TRUE
  ),
 name = "Rotate",
  group_cols = "G"
# Add the `cluster_group` transformation
# Note that this function requires the entire input data
# to properly scale the groups. We therefore specify `group_cols`
# as part of `args`. This works as `cluster_groups()` accepts that
# argument.
pipe$add_transformation(
  fn = cluster_groups,
  args = list(
```

74 position\_max

```
cols = c("Index", "A"),
    suffix = "",
    overwrite = TRUE,
    multiplier = 0.05,
    group_cols = "G"
    ),
    name = "Cluster"
)

# Check pipeline object
pipe

# Apply pipeline to data.frame
# Enable `verbose` to print progress
pipe$apply(df, verbose = TRUE)
```

position\_max

Positions the highest values with values decreasing around it

## **Description**

## [Experimental]

The highest value is positioned at the given index/quantile with the other values decreasing around it.

## Example:

```
The column values:
```

```
c(1, 2, 3, 4, 5)
and position = 2
are ordered as:
c(3, 5, 4, 2, 1)
```

#### **Usage**

```
position_max(data, col = NULL, position = NULL, shuffle_sides = FALSE)
```

### **Arguments**

data data.frame or vector.

col Column to create sorting factor by. When `NULL` and `data` is a data.frame,

the row numbers are used.

position Index or quantile (in  $\theta$ -1) at which to position the element of interest.

shuffle\_sides Whether to shuffle which elements are left and right of the position. (Logical)

### Value

The sorted data.frame (tibble) / vector.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

position\_min 75

#### See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), pair_extremes(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "index" = 1:10,
  "A" = sample(1:10),
  "B" = runif(10),
  "C" = LETTERS[1:10],
  G'' = c(
   1, 1, 1, 2, 2,
    2, 3, 3, 3, 3
  ),
  stringsAsFactors = FALSE
# Position the highest index (row number)
position_max(df, position = 3)sindex
position_max(df, position = 8)$index
# Position the maximum value in each of the columns
position_max(df, col = "A", position = 3)$A
position_max(df, col = "B", position = 3)$B
position_max(df, col = "C", position = 3)$C
# Randomize which elements are left and right of the position
position_max(df, col = "A", position = 3, shuffle_sides = TRUE)$A
# Grouped by G
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  position_max(col = "A", position = 2)
# Plot the rearranged values
plot(x = 1:10, y = position_max(df, col = "B", position = 3)$B)
plot(x = 1:10, y = position_max(df, col = "B", position = 3, shuffle_sides = TRUE)$B)
```

76 position\_min

## **Description**

# [Experimental]

The lowest value is positioned at the given index/quantile with the other values increasing around it

## Example:

```
The column values:
```

```
c(1, 2, 3, 4, 5)
and position = 2
are ordered as:
```

# c(3, 1, 2, 4, 5)

## Usage

```
position_min(data, col = NULL, position = NULL, shuffle_sides = FALSE)
```

# **Arguments**

data data.frame or vector.

col Column to create sorting factor by. When `NULL` and `data` is a data. frame,

the row numbers are used.

position Index or quantile (in 0-1) at which to position the element of interest.

shuffle\_sides Whether to shuffle which elements are left and right of the position. (Logical)

## Value

The sorted data.frame (tibble) / vector.

# Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

## See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), pair_extremes(), position_max(), rev_windows(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)

# Set seed
set.seed(1)

# Create a data frame
df <- data.frame(
   "index" = 1:10,
   "A" = sample(1:10),
   "B" = runif(10),</pre>
```

rev\_windows 77

```
C'' = LETTERS[1:10],
  G'' = c(
   1, 1, 1, 2, 2,
    2, 3, 3, 3, 3
  stringsAsFactors = FALSE
# Position the smallest index (row number)
position_min(df, position = 3)$index
position_min(df, position = 8)$index
# Position the minimum value in each of the columns
position_min(df, col = "A", position = 3)$A
position_min(df, col = "B", position = 3)$B
position_min(df, col = "C", position = 3)$C
# Randomize which elements are left and right of the position
position\_min(df, col = "A", position = 3, shuffle\_sides = TRUE)$A
# Grouped by G
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  position_min(col = "A", position = 2)
# Plot the rearranged values
plot(x = 1:10, y = position_min(df, col = "B", position = 3)$B)
plot(x = 1:10, y = position_min(df, col = "B", position = 3, shuffle_sides = TRUE)$B)
```

rev\_windows

Reverse order window-wise

### **Description**

### [Experimental]

The values are windowed and reversed within windows.

The \*\_vec() version takes and returns a vector.

## Example:

```
The column values:
```

```
c(1, 2, 3, 4, 5, 6)
```

With window\_size = 3

# Are ordered as:

```
c(3, 2, 1, 6, 4, 5)
```

# Usage

```
rev_windows(data, window_size, factor_name = ".window", overwrite = FALSE)
rev_windows_vec(data, window_size)
```

78 rev\_windows

#### **Arguments**

data data.frame or vector.

window\_size Size of the windows. (Logical)

factor\_name Name of the factor with window identifiers. If `NULL`, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### Value

The sorted data.frame (tibble) / vector. Optionally with the windows factor added.

When `data` is a vector and `keep\_windows` is FALSE, the output will be a vector. Otherwise, a data. frame.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), pair_extremes(), position_max(), position_min(), roll_elements(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "index" = 1:10,
  "A" = sample(1:10),
  "B" = runif(10),
  "C" = LETTERS[1:10],
  "G" = rep(1:2, each = 5),
  stringsAsFactors = FALSE
# For vector
rev_windows_vec(1:10, window_size = 3)
# For data frame
rev_windows(df, window_size = 3)
rev_windows(df, window_size = 3, factor_name = NULL)
# Grouped by G
df %>%
  dplyr::select(G, index) %>% # For clarity
  dplyr::group_by(G) %>%
  rev_windows(window_size = 3)
```

roll\_elements 79

```
# Plot the extreme pairs
plot(
    x = 1:10,
    y = rev_windows_vec(1:10, window_size = 3)
)
```

roll\_elements

Roll elements

# Description

# [Experimental]

Rolls positions of elements.

Example:

```
Rolling c(1, 2, 3, 4, 5) with `n = 2` becomes:
c(3, 4, 5, 1, 2)
```

roll\_elements\_vec() takes and returns a vector.

Should not be confused with roll\_values(), which changes the *values* of the elements and wraps to a given range.

# Usage

```
roll_elements(
  data,
  cols = NULL,
  n = NULL,
  n_fn = NULL,
  n_col_name = ".n",
  overwrite = FALSE,
  ...
)
roll_elements_vec(data, n = NULL, n_fn = NULL, ...)
```

# **Arguments**

data	$\hbox{vector or data.frame to roll elements of. When a {\tt data.frame} is grouped, the rolling is applied group-wise.}\\$
cols	Names of columns in `data` to roll. If NULL, the $index$ is rolled and used to reorder `data`.
	<b>N.B.</b> only used when `data` is a data.frame.
n	Number of positions to roll. A positive number rolls <i>left/up</i> . A negative number rolls <i>right/down</i> .
n_fn	Function to find `n`. Useful when `data` is a grouped data.frame and `n` should depend on the rows in the group.
	<b>Input</b> : Each specified vector/column in `data` is passed to the function as a separate argument

separate argument.

Output: It should return either a vector with one integer-like scalar ner

**Output**: It should return either a vector with one integer-like scalar *per column* or a single integer-like scalar to use for all columns.

Can be created with create\_n\_fn(). See also median\_index() and quantile\_index().

80 roll\_elements

```
n_col_name Name of new column with the applied `n` values. If NULL, no column is added.

overwrite Whether to allow overwriting of columns with the same name as `n_col_name`.

(Logical)

Extra arguments for `n_fn`.
```

#### Value

Rolled `data`.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

```
Other roll functions: roll_values()
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), pair_extremes(), position_max(), position_min(), rev_windows(), shuffle_hierarchy(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)
# Roll vector left
roll_elements(1:10, n = 2)
# Roll vector right and return the vector
roll_elements_vec(1:10, n = -2)
# Roll vector left by median index (rounded to 6)
roll_elements(3:12, n_fn = median_index)
# Roll vector right by median value (rounded to 8)
roll_elements(3:12, n_fn = create_n_fn(median, negate = TRUE))
# Pass extra arguments (here 'prob') to 'n_fn' via '...'
roll_elements(
  1:10,
 n_fn = quantile_index,
  prob = 0.2
# Roll data.frame
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  x'' = 1:20,
```

roll\_values 81

```
"y" = runif(20) * 10,
  "g" = rep(1:4, each = 5)
)

# Roll rows left/up
roll_elements(df, n = 2)

# Roll rows right/down
roll_elements(df, n = -2)

# Roll 'x' column right/down
roll_elements(df, cols = "x", n = -2)

# Roll rows right by median index in each group
# Specify 'negate' for the 'median_index' function
roll_elements(
   df %>% dplyr::group_by(g),
   n_fn = median_index,
   negate = TRUE
)
```

roll\_values

Shift values and wrap to range

## **Description**

### [Experimental]

Adds a specified value to each element in the vector and wraps the values around the min-max range with:

```
(x - .min) \% (.max - .min + between) + .min
```

Useful when adding to the degrees of a circle, where the values should remain in the 0-360 range. A value larger than 360 will start over from 0, e.g. 365->5, while a value smaller than 0 would become e.g. -5->355. Here, 0 and 360 are considered the same angle. If we were instead adding days to the weekdays 1-7, where 1 and 7 are separate days, we can set `between = 1` to have one day in-between them.

```
wrap_to_range() is a wrapper with `add = 0`.
```

The \*\_vec() versions take and return a vector.

Should not be confused with roll\_elements(), which changes the *positions* of the elements.

#### Usage

```
roll_values(
  data,
  cols = NULL,
  add = 0,
  .min = NULL,
  .max = NULL,
  between = 0,
  na.rm = FALSE,
  suffix = "_rolled",
  keep_original = TRUE,
```

82 roll\_values

```
range_col_name = ".range",
      overwrite = FALSE
    wrap_to_range(
      data,
      cols = NULL,
      .min = NULL,
      .max = NULL,
      between = 0,
      na.rm = FALSE,
      suffix = "_wrapped",
      keep_original = TRUE,
      range_col_name = ".range",
      overwrite = FALSE
    roll_values_vec(
      data,
      add = 0,
      .min = NULL,
      .max = NULL,
      between = 0,
      na.rm = FALSE
   wrap_to_range_vec(data, .min = NULL, .max = NULL, between = 0, na.rm = FALSE)
Arguments
    data
                     vector or data. frame to roll/wrap values of. When a data. frame is grouped,
                     the rolling/wrapping is applied group-wise.
                     Names of columns to roll/wrap in `data`. Must be specified when `data` is a
    cols
                     data.frame.
                     Amount to add to each element. (numeric scalar)
    add
                     When 0, the wrapping is applied without any rolling.
                     Minimum value allowed. If NULL, the minimum value in the vector/column is
    .min
                     used.
                     Maximum value allowed. If NULL, the maximum value in the vector/column is
    . max
    between
                     The wrapping distance between `.max` and `.min`.
                     When 0, they are considered the same. I.e. '.max == .min'.
                     When 1, `x` can be greater than `.max` by up to 1, why `.min` and `.max`
                     are two separate values with 1 in-between them. I.e. '.max + 1 == .min'.
                     Whether to remove missing values (NAs) when finding the `.min` and `.max`
    na.rm
                     values.
    suffix
                     Suffix to add to the names of the generated columns.
                     Use an empty string (i.e. "") to overwrite the original columns.
                     Whether to keep the original columns. (Logical)
    keep_original
```

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

roll\_values 83

```
range_col_name Name of new column with the min-max range. If NULL, no column is added.

N.B. Ignored when `data` is a vector.

overwrite Whether to allow overwriting of existing columns. (Logical)
```

#### Value

`data` with new columns with values in the specified min-max range(s) and columns with the applied ranges.

The \*\_vec() versions return a vector.

### Author(s)

```
Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>
```

#### See Also

```
Other roll functions: roll_elements()
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()
```

```
# Attach packages
library(rearrr)
# Add 90 to all degrees
# Note that 0 and 360 is considered the same angle
# why there is no distance between the two
roll_values(c(0:360), add = 90)
# Get as vector
roll_values_vec(c(0:360), add = 90)
# Change limits to 0-180
# so e.g. 270 becomes 90
roll_values(c(0:360), .min = 0, .max = 180)
# Change values first, then wrap to range
x < -c(1:7)
x <- x^2
wrap_to_range(x, .min = 1, .max = 7)
# With 1 in-between .min and .max
wrap_{to} = 1, .max = 7, between = 1
# Get as vector
wrap_{to} = 1, .max = 7, between = 1)
# Roll data.frame
# Set seed
set.seed(1)
```

84 rotate\_2d

```
# Create a data frame
df <- data.frame(</pre>
  w'' = 1:7,
  "d" = c(0, 45, 90, 135, 180, 270, 360)
# Roll weekdays by 1 day
roll_values(
  df,
 cols = "w",
 add = 1,
  .min = 1,
  .max = 7,
 between = 1
)
# Roll degrees by -90 degrees
roll_values(
  df,
  cols = "d"
  add = -90,
  .min = 0,
  .max = 360,
 between = 0
# Roll both weekdays and degrees by 1
# We don't specify .min and .max, so they
# are based on the values in the columns
# Note: This is not that meaningful but shows
# the option
roll_values(
  df,
  cols = c("w", "d"),
  add = 1
)
# Wrap weekdays to 2-5 range
wrap_to_range(
  df,
 cols = "w",
  .min = 2,
  .max = 5,
  between = 1
```

rotate\_2d

Rotate the values around an origin in 2 dimensions

# Description

# [Experimental]

The values are rotated counterclockwise around a specified origin.

rotate\_2d 85

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and rotating around e.g. the centroid of each group.

# Usage

```
rotate_2d(
  data,
  degrees,
  x_col = NULL,
  y_col = NULL,
  suffix = "_rotated",
  origin = NULL,
  origin_fn = NULL,
  keep_original = TRUE,
  degrees_col_name = ".degrees",
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

# **Arguments**

origin\_col\_name

overwrite

data	data.frame or vector.		
degrees	Degrees to rotate values counterclockwise. In $[-360, 360]$ . Can be a vector with multiple degrees.		
x_col	Name of x column in `data`. If NULL and `data` is a vector, the index of `data` is used. If `data` is a data.frame, it must be specified.		
y_col	Name of y column in `data` . If `data` is a data . frame, it must be specified.		
suffix	Suffix to add to the names of the generated columns.		
	Use an empty string (i.e. "") to overwrite the original columns.		
origin	Coordinates of the origin to rotate around. A vector with 2 elements (i.e. origin_x, origin_y). Ignored when `origin_fn` is not NULL.		
origin_fn	Function for finding the origin coordinates.		
	<b>Input</b> : Each column will be passed as a vector in the order of `cols`.		
	Output: A vector with one scalar per dimension.		
	Can be created with create_origin_fn() if you want to apply the same function to each dimension.		
	E.g. `create_origin_fn(median)` would find the median of each column.		
	<pre>Built-in functions are centroid(), most_centered(), and midrange()</pre>		
keep_original	Whether to keep the original columns. (Logical)		
	Some columns may have been overwritten, in which case only the newest versions are returned.		
degrees_col_name			
	Name of new column with the degrees. If NULL, no column is added.		

Name of new column with the origin coordinates. If NULL, no column is added.

Whether to allow overwriting of existing columns. (Logical)

86 rotate\_2d

#### **Details**

Applies the following rotation matrix:

```
\begin{bmatrix} \cos\theta & , -\sin\theta \end{bmatrix} \begin{bmatrix} \sin\theta & , \cos\theta \end{bmatrix}
```

```
That is:
```

```
x' = x\cos\theta - y\sin\thetay' = x\sin\theta + y\cos\theta
```

Where  $\theta$  is the angle in radians.

As specified at Wikipedia/Rotation\_matrix.

### Value

data.frame (tibble) with seven new columns containing the rotated x-,y- and z-values and the degrees, radiuses and origin coordinates.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()

Other rotation functions: rotate_3d(), swirl_2d(), swirl_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "Index" = 1:12,
  A'' = c(1:4, 9:12, 15:18),
  "G" = rep(1:3, each = 4)
# Rotate values around (0, 0)
rotate_2d(df, degrees = 45, x_col = "Index", y_col = "A", origin = c(0, 0))
# Rotate A around the centroid
df_rotated <- df %>%
  rotate_2d(
    x_col = "Index",
```

```
y_{col} = "A",
    degrees = c(0, 120, 240),
    origin_fn = centroid
df_rotated
# Plot A and A rotated around overall centroid
ggplot(df_rotated, aes(x = Index_rotated, y = A_rotated, color = factor(.degrees))) +
  geom_hline(yintercept = mean(df$A), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_vline(xintercept = mean(df$Index), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_line(alpha = .4) +
  geom_point() +
  theme_minimal() +
  labs(x = "Index", y = "Value", color = "Degrees")
# Rotate around group centroids
df_grouped <- df %>%
  dplyr::group_by(G) %>%
  rotate_2d(
   x_{col} = "Index",
    y_col = "A",
    degrees = c(0, 120, 240),
    origin_fn = centroid
df_grouped
# Plot A and A rotated around group centroids
ggplot(df_grouped, aes(x = Index_rotated, y = A_rotated, color = factor(.degrees))) +
  geom_point() +
  theme_minimal() +
  labs(x = "Index", y = "Value", color = "Degrees")
```

rotate\_3d

Rotate the values around an origin in 3 dimensions

# **Description**

# [Experimental]

The values are rotated counterclockwise around a specified origin.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and rotating around e.g. the centroid of each group.

## Usage

```
rotate_3d(
   data,
   x_col,
   y_col,
   z_col,
   x_deg = 0,
   y_deg = 0,
   z_deg = 0,
   suffix = "_rotated",
```

```
origin = NULL,
origin_fn = NULL,
keep_original = TRUE,
degrees_col_name = ".degrees",
origin_col_name = ".origin",
overwrite = FALSE
)
```

#### **Arguments**

```
data data.frame or vector. x\_col, y\_col, z\_col
```

Name of x/y/z column in `data`. All must be specified.

x\_deg, y\_deg, z\_deg

Degrees to rotate values around the x/y/z-axis counterclockwise. In [-360, 360]. Can be vectors with multiple degrees.

`x\_deg` is roll. `y\_deg` is pitch. `z\_deg` is yaw.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

origin Coordinates of the origin to rotate around. Vector with 3 elements (i.e. ori-

gin\_x, origin\_y, origin\_z). Ignored when `origin\_fn` is not NULL.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

**Output**: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

 $E.g. \ \verb|`create_origin_fn(median)| \ would find the median of each column.$ 

**Built-in functions** are centroid(), most\_centered(), and midrange()

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest versions are returned.

degrees\_col\_name

Name of new column with the degrees. If NULL, no column is added.

Also adds a string version with the same name + "\_str", making it easier to group by the degrees when plotting multiple rotations.

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

### **Details**

Applies the following rotation matrix:

Where  $\alpha = `z_{deg}`$  in radians,  $\beta = `y_{deg}`$  in radians,  $\gamma = `x_{deg}`$  in radians.

As specified at Wikipedia/Rotation\_matrix.

#### Value

data.frame (tibble) with six new columns containing the rotated x-,y- and z-values and the degrees and origin coordinates.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), shear_2d(), shear_3d(), swirl_2d(), swirl_3d()

Other rotation functions: rotate_2d(), swirl_2d(), swirl_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Set seed
set.seed(3)
# Create a data frame
df <- data.frame(</pre>
  "x" = 1:12,
  y'' = c(1:4, 9:12, 15:18),
  z" = runif(12),
  "g" = rep(1:3, each = 4)
# Rotate values 45 degrees around x-axis at (0, 0, 0)
rotate_3d(df, x_col = "x", y_col = "y", z_col = "z", x_deg = 45, origin = c(0, 0, 0))
# Rotate all axes around the centroid
df_rotated <- df %>%
  rotate_3d(
    x_{col} = "x",
    y_{col} = "y",
    z_{col} = "z"
    x_{deg} = c(0, 72, 144, 216, 288),
    y_{deg} = c(0, 72, 144, 216, 288),
    z_{deg} = c(0, 72, 144, 216, 288),
    origin_fn = centroid
 )
df_rotated
# Plot rotations
ggplot(df_rotated, aes(x = x_rotated, y = y_rotated, color = .degrees_str, alpha = z_rotated)) +
  geom_vline(xintercept = mean(df$x), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_hline(yintercept = mean(df$y), size = 0.2, alpha = .4, linetype = "dashed") +
  geom\_line(alpha = .4) +
  geom_point() +
```

```
theme_minimal() +
  labs(x = "x", y = "y", color = "degrees", alpha = "z (opacity)")
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
  x = df_rotated x_rotated
 y = df_rotated$y_rotated,
 z = df_rotated$z_rotated,
 type = "scatter3d",
 mode = "markers",
  color = df_rotated$.degrees_str
## End(Not run)
# Rotate randomly around all axes
df_rotated <- df %>%
  rotate_3d(
   x_{col} = "x"
   y_{col} = "y",
    z_{col} = "z"
   x_{deg} = round(runif(10, min = 0, max = 360)),
   y_{deg} = round(runif(10, min = 0, max = 360)),
   z_{deg} = round(runif(10, min = 0, max = 360)),
   origin_fn = centroid
 )
df_rotated
# Plot rotations
ggplot(df_rotated, aes(x = x_rotated, y = y_rotated, color = .degrees_str, alpha = z_rotated)) +
  geom_vline(xintercept = mean(df$x), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_hline(yintercept = mean(df$y), size = 0.2, alpha = .4, linetype = "dashed") +
  geom\_line(alpha = .4) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "degrees", alpha = "z (opacity)")
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
  x = df_rotated$x_rotated,
  y = df_rotated$y_rotated,
  z = df_rotated$z_rotated,
  type = "scatter3d",
 mode = "markers",
  color = df_rotated$.degrees_str
## End(Not run)
# Rotate around group centroids
df_grouped <- df %>%
  dplyr::group_by(g) %>%
```

shear\_2d 91

```
rotate_3d(
    x_{col} = "x"
    y_{col} = "y",
    z_{col} = z''
    x_{deg} = c(0, 72, 144, 216, 288),
    y_{deg} = c(0, 72, 144, 216, 288),
    z_{\text{deg}} = c(0, 72, 144, 216, 288),
    origin_fn = centroid
  )
# Plot A and A rotated around group centroids
ggplot(df_grouped, aes(x = x_rotated, y = y_rotated, color = .degrees_str, alpha = z_rotated)) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "degrees", alpha = "z (opacity)")
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
  x = df_grouped$x_rotated,
  y = df_grouped$y_rotated,
  z = df_grouped$z_rotated,
  type = "scatter3d",
 mode = "markers",
  color = df_grouped$.degrees_str
## End(Not run)
```

shear\_2d

Shear the values around an origin in 2 dimensions

## **Description**

## [Experimental]

Shear a set of 2d points around an origin. The shearing formulas (excluding the origin movements) is:

$$x' = x + x_s hear * y$$
$$y' = y + y_s hear * x$$

The data points in `data` are moved prior to the shearing, to bring the origin to 0 in all dimensions. After the shearing, the inverse move is applied to bring the origin back to its original position.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and shearing around e.g. the centroid of each group.

### Usage

```
shear_2d(
  data,
  x_shear,
  y_shear = 0,
```

92 shear\_2d

```
x_col = NULL,
y_col = NULL,
suffix = "_sheared",
origin = NULL,
origin_fn = NULL,
keep_original = TRUE,
shear_col_name = ".shear",
origin_col_name = ".origin",
overwrite = FALSE
)
```

# **Arguments**

data	data.frame or vector.	
x_shear	Shear factor for the $x$ dimension (numeric). Decides the amount of shearing. Can be a vector with multiple shear factors.	
y_shear	Shear factor for the y dimension (numeric). Decides the amount of shearing. Can be a vector with multiple shear factors.	
x_col	Name of x column in `data`.	
y_col	Name of y column in `data`.	
suffix	Suffix to add to the names of the generated columns.	
	Use an empty string (i.e. "") to overwrite the original columns.	
origin	Coordinates of the origin to shear around. Vector with 2 elements (origin $_x$ , origin $_y$ ). Ignored when `origin $_f$ n` is not NULL.	
origin_fn	Function for finding the origin coordinates.	
	<b>Input</b> : Each column will be passed as a vector in the order of `cols`.	
	Output: A vector with one scalar per dimension.	
	Can be created with create_origin_fn() if you want to apply the same function to each dimension.	
	E.g. `create_origin_fn(median)` would find the median of each column.	
	<pre>Built-in functions are centroid(), most_centered(), and midrange()</pre>	
keep_original	Whether to keep the original columns. (Logical)	
	Some columns may have been overwritten, in which case only the newest versions are returned.	
shear_col_name	Name of new column with the shearing factors. If NULL, no column is added.	
	Also adds a string version with the same name + "_str", making it easier to group by the shearing factors when plotting multiple shearings.	
origin_col_name		
	Name of new column with the origin coordinates. If NULL, no column is added.	
overwrite	Whether to allow overwriting of existing columns. (Logical)	

# Value

 ${\tt data.frame}$  (tibble) with sheared columns, the shearing factors and the origin coordinates.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

shear\_2d 93

#### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_3d(), swirl_2d(), swirl_3d()

Other shearing functions: shear_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Create a data frame
df <- data.frame(</pre>
  x'' = rep(1:6, each = 2),
  "y" = rep(c(1, 4), 6),
  "g" = rep(1:2, each = 6)
# Shear the x variable with regards to y
# around the centroid
df_sheared <- shear_2d(</pre>
  data = df,
 x_{shear} = 2.5,
 x_{col} = "x",
 y_{col} = "y",
 origin_fn = centroid
)
# Plot sheared data
# Black: original points
# Red: sheared points
df_sheared %>%
  ggplot(aes(x = x, y = y)) +
  geom_point() +
  geom_point(aes(x = x_sheared, y = y_sheared, color = "red")) +
  theme_minimal()
# Shear in both dimensions
df_sheared <- shear_2d(
 data = df,
 x_shear = 2.5,
 y_shear = 2.5,
 x_{col} = "x",
 y_{col} = "y",
  origin_fn = centroid
# Plot sheared data
# Black: original points
# Red: sheared points
df_sheared %>%
  ggplot(aes(x = x, y = y)) +
  geom_point() +
  geom_point(aes(x = x_sheared,y = y_sheared, color = "red")) +
```

```
theme_minimal()
# Shear grouped data frame
# Affects the calculated origin
df_sheared <- shear_2d(</pre>
  data = dplyr::group_by(df, g),
  x_shear = 2.5,
 x_{col} = "x",
 y_{col} = "y",
 origin_fn = centroid
# Plot sheared data
# Black: original points
# Red: sheared points
df_sheared %>%
  ggplot(aes(x = x, y = y)) +
  geom_point() +
  geom_point(aes(x = x_sheared, y = y_sheared, color = "red")) +
  theme_minimal()
# Shear a vector with multiple shearing factors
shear_2d(
  data = c(1:10),
  x_{shear} = c(1, 1.5, 2, 2.5),
  origin = c(0, 0)
```

shear\_3d

Shear values around an origin in 3 dimensions

### **Description**

## [Experimental]

Shears points around an origin in 3-dimensional space. See applied shearing matrices under **Details**.

The data points in `data` are moved prior to the shearing, to bring the origin to 0 in all dimensions. After the shearing, the inverse move is applied to bring the origin back to its original position.

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and shearing around e.g. the centroid of each group.

### Usage

```
shear_3d(
  data,
  x_col,
  y_col,
  z_col,
  x_shear = NULL,
  y_shear = NULL,
  z_shear = NULL,
  suffix = "_sheared",
  origin = NULL,
```

shear\_3d 95

```
origin_fn = NULL,
keep_original = TRUE,
shear_col_name = ".shear",
origin_col_name = ".origin",
overwrite = FALSE
)
```

#### **Arguments**

data data.frame or vector.  $x\_col$ ,  $y\_col$ ,  $z\_col$ 

Name of x/y/z column in `data`. All must be specified.

x\_shear, y\_shear, z\_shear

Shear factor for the x/y/z dimension (numeric). Decides the amount of shearing. Can be vectors with multiple shear factors.

**N.B.** Exactly **2** of the dimensions must have shear factors specified.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

origin Coordinates of the origin to shear around. Vector with 3 elements (i.e. origin\_x,

origin\_y, origin\_z). Ignored when `origin\_fn` is not NULL.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

E.g. `create\_origin\_fn(median)` would find the median of each column.

**Built-in functions** are centroid(), most\_centered(), and midrange()

keep\_original Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest ver-

sions are returned.

shear\_col\_name Name of new column with the shearing amounts. If NULL, no column is added.

Also adds a string version with the same name + "\_str", making it easier to

group by the shearing amounts when plotting multiple shears.

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### **Details**

Applies one of the following transformation matrices, depending on which two shearing amounts are specified:

Given `x\_shear` and `y\_shear`:

```
[ 1 , 0 , x\_shear ]
[ 0 , 1 , y\_shear ]
[ 0 , 0 , 1 ]
```

Given `x\_shear` and `z\_shear`:

Given `y\_shear` and `z\_shear`:

### Value

data.frame (tibble) with six new columns containing the sheared x-, y- and z-values and the shearing amounts and origin coordinates.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), swirl_2d(), swirl_3d()

Other shearing functions: shear_2d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Set seed
set.seed(1)
df_square <- square(runif(100)) %>%
 rename(x = .square_x,
        y = Value) %>%
 mutate(z = 1)
# Shear the x and z axes
# around the centroid
df_sheared <- shear_3d(</pre>
 data = df_square,
  x_{col} = "x",
 y_{col} = "y",
 z_{col} = "z",
 x_shear = 2,
 z_{shear} = 4,
 origin_fn = centroid
```

shear\_3d 97

```
# Plot sheared data
# Black: original points
# Red: sheared points
df_sheared %>%
  ggplot(aes(x = x, y = y)) +
  geom_point() +
  geom_point(aes(x = x_sheared, y = y_sheared, color = "red")) +
  theme_minimal()
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
  x = df\_sheared$x\_sheared,
 y = df_sheared$y_sheared,
  z = df_sheared$z_sheared,
 type = "scatter3d",
 mode = "markers",
  color = df_sheared$.shear_str
## End(Not run)
# Shear the y and z axes
# around the centroid
df_sheared <- shear_3d(</pre>
 data = df_square,
  x_{col} = "x",
  y_{col} = "y",
 z_{col} = "z".
 y_shear = 2,
 z_{shear} = 4,
 origin_fn = centroid
# Plot sheared data
# Black: original points
# Red: sheared points
df_sheared %>%
  ggplot(aes(x = x, y = y)) +
  geom_point() +
  geom_point(aes(x = x_sheared, y = y_sheared, color = "red")) +
  theme_minimal()
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
  x = df\_sheared$x\_sheared,
  y = df_sheared$y_sheared,
 z = df_sheared$z_sheared,
 type = "scatter3d",
 mode = "markers",
 color = df_sheared$.shear_str
## End(Not run)
\# Shear the y and z axes with multiple amounts at once
```

98 shuffle\_hierarchy

```
# around the centroid
df_sheared <- shear_3d(</pre>
  data = df_square,
  x_{col} = "x",
 y_col = "y"
 z_{col} = z''
 y_{shear} = c(0, 2, 4),
 z_{shear} = c(0, 4, 6),
 origin_fn = centroid
# Plot sheared data
df_sheared %>%
  ggplot(aes(x = x\_sheared, y = y\_sheared, color = .shear\_str)) +
  geom_point() +
  theme_minimal()
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
 x = df_sheared$x_sheared,
  y = df_sheared$y_sheared,
  z = df_sheared$z_sheared,
  type = "scatter3d",
 mode = "markers",
  color = df_sheared$.shear_str
## End(Not run)
```

shuffle\_hierarchy

Shuffle multi-column hierarchy of groups

## **Description**

## [Experimental]

Shuffles a tree/hierarchy of groups, one column at a time. The levels in the last ("leaf") column are shuffled first, then the second-last column, and so on. Elements of the same group are ordered sequentially.

## Usage

```
shuffle_hierarchy(
  data,
  group_cols,
  cols_to_shuffle = group_cols,
  leaf_has_groups = TRUE
)
```

## **Arguments**

data data.frame.

shuffle\_hierarchy 99

```
group_cols Names of columns making up the group hierarchy. The last column is the leaf and is shuffled first (if also in `cols_to_shuffle`).
```

cols\_to\_shuffle

Names of columns to shuffle hierarchically. By default, all the `group\_cols` are shuffled.

leaf\_has\_groups

Whether the leaf column contains groups or values. (Logical)

When the elements are *group identifiers*, they are ordered sequentially and shuffled together.

When the elements are *values*, they are simply shuffled.

#### Value

The shuffled data.frame (tibble).

# Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), pair_extremes(), position_max(), position_min(), rev_windows(), roll_elements(), triplet_extremes()
```

```
# Attach packages
library(rearrr)
library(dplyr)
df <- data.frame(</pre>
  'a' = rep(1:4, each = 4),
  b' = rep(1:8, each = 2),
  c' = 1:16
# Set seed for reproducibility
set.seed(2)
# Shuffle all columns
shuffle_hierarchy(df, group_cols = c('a', 'b', 'c'))
# Don't shuffle 'b' but keep grouping by it
# So 'c' will be shuffled within each group in 'b'
shuffle_hierarchy(
  data = df,
  group_cols = c('a', 'b', 'c'),
  cols_to_shuffle = c('a', 'c')
# Shuffle 'b' as if it's not a group column
# so elements are independent within their group
# (i.e. same-valued elements are not necessarily ordered sequentially)
shuffle_hierarchy(df, group_cols = c('a', 'b'), leaf_has_groups = FALSE)
```

100 square

square

Create x-coordinates so the points form a square

# Description

## [Experimental]

Create the x-coordinates for a vector of y-coordinates such that they form a rotated square.

This will likely look most like a square when the y-coordinates are somewhat equally distributed, e.g. a uniform distribution.

## Usage

```
square(
  data,
  y_col = NULL,
  .min = NULL,
  .max = NULL,
  offset_x = 0,
  keep_original = TRUE,
  x_col_name = ".square_x",
  edge_col_name = ".edge",
  overwrite = FALSE
)
```

# Arguments

data	data.frame or vector.
y_col	Name of column in `data` with y-coordinates to create x-coordinates for.
.min	Minimum y-coordinate. If NULL, it is inferred by the given y-coordinates.
.max	Maximum y-coordinate. If NULL, it is inferred by the given y-coordinates.
offset_x	Value to offset the x-coordinates by.
keep_original	Whether to keep the original columns. (Logical)
	Some columns may have been overwritten, in which case only the newest versions are returned.
x_col_name	Name of new column with the x-coordinates.
edge_col_name	Name of new column with the edge identifiers. If NULL, no column is added. Numbering is clockwise and starts at the upper-right edge.
overwrite	Whether to allow overwriting of existing columns. (Logical)

### Value

data.frame (tibble) with the added x-coordinates and an identifier for the edge the data point is a part of.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

square 101

### See Also

Other forming functions: circularize(), hexagonalize(), triangularize()

```
# Attach packages
library(rearrr)
library(dplyr)
library(purrr)
library(ggplot2)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
 y'' = runif(200),
  "g" = factor(rep(1:5, each = 40))
)
# Square 'y'
df_sq <- square(df, y_col = "y")</pre>
df_sq
# Plot square
df_sq %>%
  ggplot(aes(x = .square_x, y = y, color = .edge)) +
  geom_point() +
  theme_minimal()
# Grouped squaring
# Square 'y' for each group
# First cluster the groups a bit to move the
# squares away from each other
df_sq \leftarrow df \%
 cluster_groups(
    cols = "y",
    group_cols = "g",
    suffix = "",
    overwrite = TRUE
  ) %>%
  dplyr::group_by(g) %>%
  square(
    y_{col} = "y",
    overwrite = TRUE
# Plot squares
df_sq %>%
  ggplot(aes(x = .square_x, y = y, color = g)) +
  geom_point() +
  theme_minimal()
```

102 swirl\_2d

```
# Specifying minimum value
# Specify minimum value manually
df_sq \leftarrow square(df, y_col = "y", .min = -2)
df_sq
# Plot square
df_sq %>%
  ggplot(aes(x = .square_x, y = y, color = .edge)) +
  geom_point() +
  theme_minimal()
# Multiple squares by contraction
# Start by squaring 'y'
df_sq \leftarrow square(df, y_col = "y")
# Contract '.square_x' and 'y' towards the centroid
# To contract with multiple multipliers at once,
# we wrap the call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  x = c(1, 0.75, 0.5, 0.25, 0.125),
  .f = function(mult) {
    expand_distances(
      data = df_sq,
      cols = c(".square_x", "y"),
      multiplier = mult,
      origin_fn = centroid,
      overwrite = TRUE
  }
)
df_expanded
df_expanded %>%
  ggplot(aes(
    x = .square_x_expanded, y = y_expanded,
    color = .edge, alpha = .multiplier
  )) +
  geom_point() +
  theme_minimal()
```

swirl\_2d

Swirl the values around an origin in 2 dimensions

### **Description**

### [Experimental]

The values are swirled counterclockwise around a specified origin. The swirling is done by rotating around the origin with the degrees based on the distances to the origin as so:

```
degrees = scale_f n(distances)/(2 * radius) * 360
```

swirl\_2d 103

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and swirling around e.g. the centroid of each group.

### Usage

```
swirl_2d(
 data,
  radius,
 x_{col} = NULL
 y_{col} = NULL,
 suffix = "_swirled",
 origin = NULL,
 origin_fn = NULL,
  scale_fn = identity,
 keep_original = TRUE,
 degrees_col_name = ".degrees",
 radius_col_name = ".radius",
 origin_col_name = ".origin",
 overwrite = FALSE
```

## **Arguments**

data data.frame or vector.

radius Radius of the most-inner swirl on the x-axis in the *simplest* case. A negative

number changes the direction to clockwise rotation. Can be a vector with multiple radiuses.

Note: With a custom `scaling\_fn`, this might not be the actual swirl radius anymore. Think of it more as a width setting where a larger number leads to

fewer full rotations.

Name of x column in `data`. If NULL and `data` is a vector, the index of x\_col

'data' is used. If 'data' is a data. frame, it must be specified.

y\_col Name of y column in `data`. If `data` is a data. frame, it must be specified.

Suffix to add to the names of the generated columns. suffix

Use an empty string (i.e. "") to overwrite the original columns.

Coordinates of the origin to swirl around. Vector with 2 elements (i.e. origin x, origin

origin\_y). Ignored when `origin\_fn` is not NULL.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

E.g. `create\_origin\_fn(median)` would find the median of each column.

**Built-in functions** are centroid(), most\_centered(), and midrange()

scale\_fn Function for scaling the distances before calculating the degrees.

Input: A numeric vector (the distances).

**Output**: A numeric vector (the scaled distances) of the same length.

E.g.:

function(d){

104 swirl\_2d

#### Value

data.frame (tibble) with three new columns containing the swirled x- and y-values, the degrees, the radius, and the origin coordinates.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_3d()

Other rotation functions: rotate_2d(), rotate_3d(), swirl_3d()

Other distance functions: closest_to(), dim_values(), distance(), expand_distances_each(), expand_distances(), furthest_from(), swirl_3d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
# Set seed
set.seed(4)
# Create a data frame
df <- data.frame(</pre>
  "x" = 1:50,
  y'' = 1,
  "r1" = runif(50),
  "r2" = runif(50) * 35,
  g'' = rep(1:5, each = 10)
# Swirl values around (0, 0)
swirl_2d(
  data = df,
```

```
radius = 45,
 x_{col} = "x",
 y_col = "y",
 origin = c(0, 0)
# Swirl around the centroid
# with 6 different radius settings
# Scale the distances with custom function
df_swirled <- swirl_2d(</pre>
 data = df,
 radius = c(95, 96, 97, 98, 99, 100),
 x_{col} = "x",
 y_col = "y",
 origin_fn = centroid,
 scale_fn = function(d) {
   d^1.6
 }
)
df_swirled
# Plot swirls
df_swirled %>%
  ggplot(aes(x = x_swirled, y = y_swirled, color = factor(.radius))) +
  geom_point() +
 theme_minimal() +
 labs(x = "x", y = "y", color = ".radius")
# Swirl random data
# The trick lies in finding the right radius
# Swirl the random columns
df_swirled <- swirl_2d(</pre>
 data = df,
 radius = 5,
 x_{col} = "r1",
 y_{col} = "r2",
 origin_fn = centroid
# Plot swirls
df_swirled %>%
  ggplot(aes(x = r1\_swirled, y = r2\_swirled)) +
  geom_point() +
  theme_minimal() +
  labs(x = "r1", y = "r2")
```

# **Description**

## [Experimental]

The values are swirled counterclockwise around a specified origin. The swirling is done by rotating around the origin, basing the degrees for each rotation-axis on the distances to the origin as so:

```
x_degrees = scale_f n(distances)/(2 * x_radius) * 360
```

The origin can be supplied as coordinates or as a function that returns coordinates. The latter can be useful when supplying a grouped data. frame and swirling around e.g. the centroid of each group.

## Usage

```
swirl_3d(
  data,
  x_col,
  y_col,
  z_col,
  x_radius = 0,
  y_radius = 0,
  z_radius = 0,
  suffix = "_swirled",
  origin = NULL,
  origin_fn = NULL,
  scale_fn = identity,
  keep_original = TRUE,
  degrees_col_name = ".degrees",
  radius_col_name = ".radius",
  origin_col_name = ".origin",
  overwrite = FALSE
)
```

### **Arguments**

```
data data.frame or vector. x_col, y_col, z_col
```

Name of x/y/z column in `data`. All must be specified.

```
x_radius, y_radius, z_radius
```

Radiuses of the most-inner swirls around each axis (in the *simplest* case). Can be vectors with multiple radiuses.

E.g. the `x\_radius` specifies the radius when rotating *around* the x-axis, not the radius *on* the x-axis.

Note: With a custom `scaling\_fn`, these might not be the actual swirl radiuses anymore. Think of them more as width settings where a larger number leads to fewer full rotations.

suffix Suffix to add to the names of the generated columns.

Use an empty string (i.e. "") to overwrite the original columns.

origin Coordinates of the origin to swirl around. Vector with 3 elements (i.e. origin\_x, origin\_y, origin\_z). Ignored when `origin\_fn` is not NULL.

origin\_fn Function for finding the origin coordinates.

**Input**: Each column will be passed as a vector in the order of `cols`.

Output: A vector with one scalar per dimension.

Can be created with create\_origin\_fn() if you want to apply the same function to each dimension.

E.g. `create\_origin\_fn(median)` would find the median of each column.

**Built-in functions** are centroid(), most\_centered(), and midrange()

scale\_fn

Function for scaling the distances before calculating the degrees.

**Input**: A numeric vector (the distances).

**Output**: A numeric vector (the scaled distances) of the same length.

E.g.:
function(d){
 d ^ 1.5
}

keep\_original

Whether to keep the original columns. (Logical)

Some columns may have been overwritten, in which case only the newest versions are returned.

degrees\_col\_name

Name of new column with the degrees. If NULL, no column is added.

Also adds a string version with the same name + "\_str", making it easier to group by the degrees when plotting multiple rotations.

radius\_col\_name

Name of new column with the radiuses. If NULL, no column is added.

origin\_col\_name

Name of new column with the origin coordinates. If NULL, no column is added.

overwrite

Whether to allow overwriting of existing columns. (Logical)

### Value

data.frame (tibble) with new columns containing the swirled x- and y-values, the degrees, the radiuses, and the origin coordinates.

## Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other mutate functions: apply_transformation_matrix(), cluster_groups(), dim_values(), expand_distances_each(), expand_distances(), flip_values(), roll_values(), rotate_2d(), rotate_3d(), shear_2d(), shear_3d(), swirl_2d()

Other rotation functions: rotate_2d(), rotate_3d(), swirl_2d()

Other distance functions: closest_to(), dim_values(), distance(), expand_distances_each(), expand_distances(), furthest_from(), swirl_2d()
```

```
# Attach packages
library(rearrr)
library(dplyr)
library(ggplot2)
```

```
# Set seed
set.seed(4)
# Create a data frame
df <- data.frame(</pre>
  "x" = 1:50,
  y'' = 1:50,
  z'' = 1:50
  "r1" = runif(50),
  "r2" = runif(50) * 35,
  "o" = 1,
  g'' = rep(1:5, each = 10)
# Swirl values around (0, 0, 0)
swirl_3d(
 data = df,
 x_radius = 45,
 x_{col} = "x",
 y_{col} = "y",
 z_{col} = "z",
 origin = c(0, 0, 0)
# Swirl around the centroid
df_swirled <- swirl_3d(</pre>
 data = df,
  x_{col} = "x"
 y_{col} = "y",
 z_{col} = "z"
  x_{radius} = c(100, 0, 0),
 y_radius = c(0, 100, 0),
 z_{radius} = c(0, 0, 100),
  origin_fn = centroid
df_swirled
# Plot swirls
ggplot(df_swirled, aes(x = x_swirled, y = y_swirled, color = .radius_str, alpha = z_swirled)) +
  geom_vline(xintercept = mean(df$x), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_hline(yintercept = mean(df$y), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_path(alpha = .4) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "radius", alpha = "z (opacity)")
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
 x = df_swirled$x_swirled,
 y = df_swirled$y_swirled,
 z = df_swirled$z_swirled,
 type = "scatter3d",
 mode = "markers",
  color = df_swirled$.radius_str
)
```

swirl\_3d 109

```
## End(Not run)
# Swirl around the centroid
df_swirled <- swirl_3d(</pre>
 data = df,
 x_{col} = "x"
 y_{col} = "y",
 z_{col} = "z".
 x_{radius} = c(50, 0, 0),
 y_radius = c(0, 50, 0),
 z_{radius} = c(0, 0, 50),
 origin_fn = centroid
df swirled
# Plot swirls
ggplot(df\_swirled, aes(x = x\_swirled, y = y\_swirled, color = .radius\_str, alpha = z\_swirled)) +
  geom_vline(xintercept = mean(df$x), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_hline(yintercept = mean(df$y), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_path(alpha = .4) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "radius", alpha = "z (opacity)")
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
 x = df_swirled$x_swirled,
 y = df_swirled$y_swirled,
 z = df_swirled$z_swirled,
 type = "scatter3d",
 mode = "markers",
  color = df_swirled$.radius_str
## End(Not run)
df_swirled <- swirl_3d(</pre>
  data = df,
  x_{col} = "x"
 y_{col} = "y",
  z_{col} = z''
  x_{radius} = c(25, 50, 25, 25),
  y_radius = c(50, 75, 100, 25),
  z_{radius} = c(75, 25, 25, 25),
  origin_fn = centroid,
  scale_fn = function(x) {
    x^0.81
  }
)
# Plot swirls
ggplot(df_swirled, aes(x = x_swirled, y = y_swirled, color = .radius_str, alpha = z_swirled)) +
  geom_vline(xintercept = mean(df$x), size = 0.2, alpha = .4, linetype = "dashed") +
```

110 swirl\_3d

```
geom_hline(yintercept = mean(df$y), size = 0.2, alpha = .4, linetype = "dashed") +
  geom_path(alpha = .4) +
  geom_point() +
  theme_minimal() +
  labs(x = "x", y = "y", color = "radius", alpha = "z (opacity)")
## Not run:
# Plot 3d with plotly
plotly::plot_ly(
 x = df_swirled$x_swirled,
 y = df_swirled$y_swirled,
 z = df_swirled$z_swirled,
 type = "scatter3d",
 mode = "markers",
 color = df_swirled$.radius_str
## End(Not run)
# Swirl random data
# The trick lies in finding the right radiuses
# Swirl the random columns
df_swirled <- swirl_3d(</pre>
 data = df,
 x_{col} = "r1"
 y_{col} = "r2",
 z_{col} = "o",
 x_{radius} = c(0, 0, 0, 0),
 y_radius = c(0, 30, 60, 90),
 z_{radius} = c(10, 10, 10, 10),
 origin_fn = centroid
)
# Not let's rotate it every 10 degrees
df_rotated <- df_swirled %>%
  rotate_3d(
   x_col = "r1_swirled",
   y_col = "r2_swirled",
   z_col = "o_swirled",
   x_{deg} = rep(0, 36),
   y_{deg} = rep(0, 36),
   z_{deg} = (1:36) * 10,
   suffix = "",
   origin = df_swirled$.origin[[1]],
   overwrite = TRUE
  )
# Plot rotated swirls
ggplot(
  df_rotated,
  aes(
   x = r1_swirled,
```

to\_unit\_length 111

```
y = r2_swirled,
color = .degrees_str,
alpha = o_swirled
)
) +
geom_vline(xintercept = mean(df$r1), size = 0.2, alpha = .4, linetype = "dashed") +
geom_hline(yintercept = mean(df$r2), size = 0.2, alpha = .4, linetype = "dashed") +
geom_point(show.legend = FALSE) +
theme_minimal() +
labs(x = "r1", y = "r2", color = "radius", alpha = "o (opacity)")
```

to\_unit\_length

Scale to unit length

### **Description**

### [Experimental]

Scales the vectors to unit length row-wise or column-wise.

The \*\_vec() version take and return a vector.

### Usage

```
to_unit_length(
  data,
  cols = NULL,
  by_row = is.data.frame(data),
  suffix = ifelse(isTRUE(by_row), "_row_unit", "_col_unit"),
  overwrite = FALSE
)

to_unit_length_vec(data)
```

### Arguments

data	data.frame or vector.
cols	Names of columns in `data` to scale.
by_row	Whether to scale row vectors instead of column vectors. (Logical)
	Note: Disable when `data` is a vector.
suffix	Suffix to add to the names of the generated columns.
	Use an empty string (i.e. "") to overwrite the original columns.
overwrite	Whether to allow overwriting of existing columns. (Logical)

#### Value

Scaled vector or data. frame (tibble) with the scaled columns.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

112 transfer\_centroids

#### See Also

Other scaling functions: min\_max\_scale()

#### **Examples**

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  x'' = runif(20),
  "y" = runif(20),
  "g" = rep(1:4, each = 5)
# Scale row-wise
to_unit_length(df, cols = c("x", "y"), by_row = TRUE)
# Scale column-wise
to_unit_length(df, cols = c("x", "y"), by_row = FALSE)
# Overwrite columns
to_unit_length(df, cols = c("x", "y"), suffix = "", overwrite = TRUE)
# By groups in 'g'
df %>%
  dplyr::group_by(g) %>%
  to_unit_length(cols = c("x", "y"), by_row = FALSE)
# Scale a vector
to_unit_length_vec(c(1:10))
to_unit_length(c(1:10), suffix = "", overwrite = TRUE)
vector_length(to_unit_length_vec(c(1:10)))
```

transfer\_centroids

Transfer centroids from one data frame to another

### Description

#### [Experimental]

Given two data. frames with the same columns (and groupings), transfer the centroids from one to the other.

This is commonly used to restore the centroids after transforming the columns.

### Usage

```
transfer_centroids(to_data, from_data, cols, group_cols = NULL)
```

transfer\_centroids 113

### **Arguments**

to\_data data.frame.

Existing `dplyr` groups are ignored. Specify in `group\_cols` instead.

from\_data data.frame with the same columns (and groupings) as `to\_data`.

Existing `dplyr` groups are ignored. Specify in `group\_cols` instead.

cols Names of numeric columns to transfer centroids to. Must exist in both `to\_data` and `from\_data`.

group\_cols

The `to\_data` data.frame (tibble) with the centroids from the `from\_data` data.frame.

#### Author(s)

Value

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

Other clustering functions: cluster\_groups(), generate\_clusters()

Names of grouping columns.

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  x'' = runif(20),
  "y" = runif(20),
  "g" = rep(1:4, each = 5)
\# Create another data frame with different x and y values
df2 <- df
df2$x <- runif(20)
df2$y <- runif(20)
# Check centroids before transfer
df %>%
  dplyr::group_by(g) %>%
  dplyr::summarize_all(mean)
df2 %>%
  dplyr::group_by(g) %>%
  dplyr::summarize_all(mean)
# Now let's transfer the centroids from df to df2
```

114 Transformation

```
df3 <- transfer_centroids(</pre>
  to_data = df2,
  from_data = df,
 cols = c("x", "y"),
  group_cols = "g"
# Check that the transfer gave us the same centroids as df
  dplyr::group_by(g) %>%
  dplyr::summarize_all(mean)
```

Transformation

Transformation

### **Description**

### [Experimental]

Container for the type of transformation used in Pipeline.

Note: For internal use.

#### **Public fields**

```
name Name of transformation.
fn Transformation function.
args list of arguments for `fn`.
group_cols Names of columns to group data. frame by before applying `fn`.
     When `NULL`, the data. frame is not grouped.
```

### Methods

### **Public methods:**

```
• Transformation$new()
```

- Transformation\$apply()
- Transformation\$print()
- Transformation\$clone()

**Method** new(): Initialize transformation.

Usage:

```
Transformation$new(fn, args, name = NULL, group_cols = NULL)
```

Arguments:

fn Transformation function.

args list of arguments for `fn`.

name Name of transformation.

group\_cols Names of columns to group data.frame by before applying `fn`. When `NULL`, the data.frame is not grouped.

**Method** apply(): Apply the transformation to a data. frame.

triangularize 115

Usage:

Transformation\$apply(data)

Arguments:

data data.frame.

A grouped data.frame will first be ungrouped. If `group\_cols` is specified, it will then be grouped by those columns.

Returns: Transformed version of `data`.

**Method** print(): Print an overview of the transformation.

Usage:

Transformation\$print(..., indent = 0, show\_class = TRUE)

Arguments:

... further arguments passed to or from other methods.

indent How many spaces to indent when printing.

show\_class Whether to print the transformation class name.

Returns: The pipeline. To allow chaining of methods.

**Method** clone(): The objects of this class are cloneable with this method.

Usage:

Transformation\$clone(deep = FALSE)

Arguments:

deep Whether to make a deep clone.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

 $Other\ transformation\ classes:\ Fixed Groups Transformation,\ Generated Transformation$ 

triangularize

Create x-coordinates so the points form a triangle

### **Description**

### [Experimental]

Create the x-coordinates for a vector of y-coordinates such that they form a triangle.

The data points are stochastically distributed based on edge lengths, why it might be preferable to set a random seed.

This will likely look most like a triangle when the y-coordinates are somewhat equally distributed, e.g. a uniform distribution.

116 triangularize

### Usage

```
triangularize(
  data,
  y_col = NULL,
  .min = NULL,
  .max = NULL,
  offset_x = 0,
  keep_original = TRUE,
  x_col_name = ".triangle_x",
  edge_col_name = ".edge",
  overwrite = FALSE
)
```

### **Arguments**

data	data.frame or vector.
y_col	Name of column in `data` with y-coordinates to create x-coordinates for.
.min	Minimum y-coordinate. If NULL, it is inferred by the given y-coordinates.
.max	Maximum y-coordinate. If NULL, it is inferred by the given y-coordinates.
offset_x	Value to offset the x-coordinates by.
keep_original	Whether to keep the original columns. (Logical)
	Some columns may have been overwritten, in which case only the newest versions are returned.
x_col_name	Name of new column with the x-coordinates.
edge_col_name	Name of new column with the edge identifiers. If NULL, no column is added.
	Numbering is clockwise and starts at the upper-right edge.
overwrite	Whether to allow overwriting of existing columns. (Logical)

## Value

data.frame (tibble) with the added x-coordinates and an identifier for the edge the data point is a part of.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

Other forming functions: circularize(), hexagonalize(), square()

```
# Attach packages
library(rearrr)
library(dplyr)
library(purrr)
library(ggplot2)
# Set seed
set.seed(1)
```

triangularize 117

```
# Create a data frame
df <- data.frame(</pre>
  "y" = runif(200),
  "g" = factor(rep(1:5, each = 40))
# Triangularize 'y'
df_tri <- triangularize(df, y_col = "y")</pre>
df_tri
# Plot triangle
df_tri %>%
  ggplot(aes(x = .triangle_x, y = y, color = .edge)) +
  geom_point() +
  theme_minimal()
# Grouped squaring
# Triangularize 'y' for each group
# First cluster the groups a bit to move the
# triangles away from each other
df_tri <- df %>%
  cluster_groups(
   cols = "y",
    group_cols = "g",
    suffix = "".
    overwrite = TRUE
  ) %>%
  dplyr::group_by(g) %>%
  triangularize(
    y_col = "y",
    overwrite = TRUE
# Plot triangles
df_tri %>%
  ggplot(aes(x = .triangle_x, y = y, color = g)) +
  geom_point() +
  theme_minimal()
# Specifying minimum value
# Specify minimum value manually
df_tri <- triangularize(df, y_col = "y", .min = -2)</pre>
df_tri
# Plot triangle
df_tri %>%
  ggplot(aes(x = .triangle_x, y = y, color = .edge)) +
  geom_point() +
  theme_minimal()
```

```
# Multiple triangles by contraction
# Start by squaring 'y'
df_tri <- triangularize(df, y_col = "y")</pre>
# Contract '.triangle_x' and 'y' towards the centroid
# To contract with multiple multipliers at once,
# we wrap the call in purrr::map_dfr
df_expanded <- purrr::map_dfr(</pre>
  .x = 1:10 / 10,
  .f = function(mult) {
    expand_distances(
      data = df_tri,
      cols = c(".triangle_x", "y"),
      multiplier = mult,
      origin_fn = centroid,
      overwrite = TRUE
  }
df_expanded
df_expanded %>%
  ggplot(aes(
    x = .triangle_x_expanded, y = y_expanded,
    color = .edge, alpha = .multiplier
  )) +
  geom_point() +
  theme_minimal()
```

triplet\_extremes

Makes triplets of extreme values and sort by them

### Description

#### [Experimental]

The values are grouped in three such that the first group is formed by the lowest and highest values and the value closest to the median, the second group is formed by the second lowest and second highest values and the value second closest to the median, and so on. The values are then sorted by these groups and their actual value.

When the number of rows/elements in `data` is not evenly divisible by three, the `unequal\_method\_1` (single excessive element) and `unequal\_method\_2` (two excessive elements) determines which element(s) should form a smaller group. This group will be the first group in a given grouping (see `num\_groupings`) with the identifier 1.

The \*\_vec() version takes and returns a vector.

### Example:

The column values:

```
c(1, 2, 3, 4, 5, 6)
```

```
Are sorted in triplets as:
```

```
c(1, 3, 6, 2, 4, 5)
```

#### Usage

```
triplet_extremes(
  data,
  col = NULL,
  middle_is = "middle",
  unequal_method_1 = "middle",
  unequal_method_2 = c("middle", "middle"),
  num_groupings = 1,
  balance = "mean",
  order_by_aggregates = FALSE,
  shuffle_members = FALSE,
  shuffle_triplets = FALSE,
  factor_name = ifelse(num_groupings == 1, ".triplet", ".tripleting"),
  overwrite = FALSE
triplet_extremes_vec(
  data,
  middle_is = "middle",
  unequal_method_1 = "middle",
  unequal_method_2 = c("middle", "middle"),
  num_groupings = 1,
  balance = "mean",
  order_by_aggregates = FALSE,
  shuffle_members = FALSE,
  shuffle_triplets = FALSE
)
```

### **Arguments**

data data.frame or vector.

col Column to create sorting factor by. When `NULL` and `data` is a data.frame, the row numbers are used.

middle\_is Whether the middle element in the triplet is the nth closest element to the median value or the nth+1 lowest/highest value.

One of: middle (default), min, or max.

Triplet grouping is performed greedily from the most extreme values to the least extreme values. E.g. c(1, 6, 12) is created before c(2, 5, 11) which is made before c(3, 7, 10).

```
When `middle_is` == 'middle', a 1:12 sequence is grouped into: c(c(1, 6, 12), c(2, 7, 11), c(3, 5, 10), c(4, 8, 9))
When `middle_is` == 'min', a 1:12 sequence is grouped into: c(c(1, 2, 12), c(3, 4, 11), c(5, 6, 10), c(7, 8, 9))
When `middle_is` == 'max', a 1:12 sequence is grouped into: c(c(1, 11, 12), c(2, 9, 10), c(3, 7, 8), c(4, 5, 6))
```

unequal\_method\_1, unequal\_method\_2

Method for dealing with either a single excessive element (`unequal\_method\_1`) or two excessive elements (`unequal\_method\_2`) when the number of rows/elements in `data` are not evenly divisible by three.

`unequal\_method\_1`: One of: min, middle or max.

`unequal\_method\_2`: Vector with two of: min, middle or max. Can be the same value twice.

Note: The excessive element(s) are extracted before triplet grouping. These elements are put in their own group and given group identifier 1.

E.g. When `unequal\_method\_2` is c("middle", "middle") the two elements closest to the median are extracted.

num\_groupings Number o

Number of times to group into triplets (recursively). At least 1.

Based on `balance`, the secondary groupings perform extreme triplet grouping on either the *sum*, *absolute difference*, *min*, or *max* of the triplet elements.

balance

What to balance triplets for in a given *secondary* triplet grouping. Either "mean", "spread", "min", or "max". Can be a single string used for all secondary groupings or one for each secondary grouping (`num\_groupings` - 1).

The first triplet grouping always groups the actual element values.

**mean:** Triplets have similar means. The values in the triplets from the previous grouping are aggregated with `sum()` and extreme triplet grouped.

**spread:** Triplets have similar spread (e.g. standard deviations). The values in the triplets from the previous triplet grouping are aggregated with `sum(abs(diff()))` and extreme triplet grouped.

min/max: Triplets have similar minimum / maximum values. The values in the triplets from the previous triplet grouping are aggregated with `min()` / `max()` and extreme triplet grouped.

order\_by\_aggregates

Whether to order the groups from initial groupings (first `num\_groupings` - 1) by their aggregate values instead of their group identifiers.

 $N.B. \ Only \ used \ when `num_groupings` > 1.$ 

shuffle\_members

Whether to shuffle the order of the group members within the groups. (Logical)

shuffle\_triplets

Whether to shuffle the order of the triplets. Triplet members remain together. (Logical)

factor\_name

Name of new column with the sorting factor. If `NULL`, no column is added.

overwrite

Whether to allow overwriting of existing columns. (Logical)

#### Value

The sorted data. frame (tibble) / vector. Optionally with the sorting factor added.

When `data` is a vector and `keep\_factors` is `FALSE`, the output will be a vector. Otherwise, a data. frame.

#### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

#### See Also

```
Other rearrange functions: center_max(), center_min(), closest_to(), furthest_from(), pair_extremes(), position_max(), position_min(), rev_windows(), roll_elements(), shuffle_hierarchy()
```

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
  "index" = 1:12,
  "A" = sample(1:12),
  "B" = runif(12),
  "C" = LETTERS[1:12],
  G'' = c(
   1, 1, 1, 1, 2, 2,
    2, 2, 3, 3, 3, 3
  stringsAsFactors = FALSE
# Triplet group extreme indices (row numbers)
triplet_extremes(df)
# Triplet group extremes in each of the columns
triplet_extremes(df, col = "A")$A
triplet_extremes(df, col = "B")$B
triplet_extremes(df, col = "C")$C
# Shuffle the members triplet-wise
# The triplets maintain their order
# but the rows within each triplet are shuffled
triplet_extremes(df, col = "A", shuffle_members = TRUE)
# Shuffle the order of the triplets
# The triplets are shuffled but
# the rows within each triplet maintain their order
triplet_extremes(df, col = "A", shuffle_triplets = TRUE)
# Use recursive grouping
# Mostly meaningful with much larger datasets
# Order initial grouping by group identifiers
triplet_extremes(df, col = "A", num_groupings = 2)
# Order initial grouping by aggregate values
triplet_extremes(df, col = "A", num_groupings = 2, order_by_aggregates = TRUE)
# Grouped by G
# Each G group only has 4 elements
# so it only creates 1 triplet and a group
# with the single excessive element
# per G group
```

122 vector\_length

```
df %>%
  dplyr::select(G, A) %>% # For clarity
  dplyr::group_by(G) %>%
  triplet_extremes(col = "A")
# Plot the extreme triplets
plot(
  x = 1:12,
 y = triplet_extremes(df, col = "A")$A,
 col = as.character(rep(1:4, each = 3))
# With shuffled triplet members (run a few times)
  x = 1:12,
 y = triplet_extremes(df, col = "A", shuffle_members = TRUE)$A,
 col = as.character(rep(1:4, each = 3))
# With shuffled triplets (run a few times)
plot(
 x = rep(1:6, each = 2),
 y = triplet_extremes(df, col = "A", shuffle_triplets = TRUE)$A,
 col = as.character(rep(1:4, each = 3))
```

vector\_length

Calculate vector length(s)

### **Description**

### [Experimental]

Calculates vector lengths/magnitudes row- or column-wise with

$$sqrt(sum(x^2))$$

Where x is the vector to get the length/magnitude of.

Should not be confused with length(), which counts the elements.

### Usage

```
vector_length(
  data,
  cols = NULL,
  by_row = is.data.frame(data),
  len_col_name = ".vec_len",
  overwrite = FALSE
)
```

### Arguments

data data.frame or vector.

cols Names of columns in `data` to measure vector length of.

by\_row Whether to measure length of row vectors instead of column vectors. (Logical)

Note: Disable when `data` is a vector.

vector\_length 123

 ${\tt len\_col\_name} \qquad {\tt Name~of~new~column~with~the~row~vector~lengths~when~`data`~is~a~data.frame}$ 

and `by\_row` is TRUE.

overwrite Whether to allow overwriting of existing columns. (Logical)

#### Value

Vector length(s).

When 'data' is a vector: scalar

When `data` is a data.frame and `by\_row` is TRUE: `data` with an extra column with row vector lengths.

When `data` is a data.frame and `by\_row` is FALSE: A data.frame with the summarized column vector lengths.

### Author(s)

Ludvig Renbo Olsen, <r-pkgs@ludvigolsen.dk>

### See Also

Other measuring functions: angle(), distance()

```
# Attach packages
library(rearrr)
library(dplyr)
# Set seed
set.seed(1)
# Create a data frame
df <- data.frame(</pre>
 x'' = runif(20),
  "y" = runif(20),
  "g" = rep(1:4, each = 5)
# Measure row-wise
vector_length(df, cols = c("x", "y"), by_row = TRUE)
# Measure column-wise
vector_length(df, cols = c("x", "y"), by_row = FALSE)
# By groups in 'g'
df %>%
  dplyr::group_by(g) %>%
  vector_length(cols = c("x", "y"), by_row = FALSE)
# Measure vector length of a vector
vector_length(c(1:10))
```

# Index

* clustering functions	rotate_2d, 84	
cluster_groups, 18	rotate_3d, 87	
generate_clusters,56	shear_2d, 91	
transfer_centroids, 112	shear_3d, 94	
* coordinate functions	swirl_2d, 102	
centroid, 11	swirl_3d, 105	
create_origin_fn, 24	* n functions	
is_most_centered, 61	create_n_fn, 22	
midrange, 64	median_index, 62	
most_centered, 66	* pipelines	
* distance functions	FixedGroupsPipeline, 41	
closest_to, 15	GeneratedPipeline, 51	
dim_values, 26	Pipeline, 71	
distance, 30	* rearrange functions	
expand_distances, 31	center_max, 7	
expand_distances_each, 36	center_min, 9	
furthest_from, 48	closest_to, 15	
swirl_2d, 102	furthest_from, 48	
swirl_3d, 105	pair_extremes, 68	
* expander functions	position_max, 74	
expand_distances, 31	position_min, 75	
expand_distances_each, 36	rev_windows, 77	
* forming functions	roll_elements, 79	
circularize, 12	shuffle_hierarchy, 98	
hexagonalize, 58	triplet_extremes, 118	
square, 100	* roll functions	
triangularize, 115	roll_elements, 79	
* function creators	roll_values, 81	
<pre>create_dimming_fn, 21</pre>	* rotation functions	
create_n_fn, 22	rotate_2d, 84	
create_origin_fn, 24	rotate_3d, 87	
* measuring functions	swirl_2d, 102	
angle, 3	swirl_3d, 105	
distance, 30	* scaling functions	
vector_length, 122	min_max_scale, 65	
* mutate functions	to_unit_length, 111	
<pre>apply_transformation_matrix, 5</pre>	* shearing functions	
cluster_groups, 18	shear_2d, 91	
dim_values, 26	shear_3d, 94	
expand_distances, 31	* transformation classes	
expand_distances_each, 36	FixedGroupsTransformation, 44	
flip_values, 46	GeneratedTransformation, 54	
roll values 81	Transformation 114	

INDEX 125

angle, 3, 31, 123 apply_transformation_matrix, 5, 19, 28, 33, 38, 47, 83, 86, 89, 93, 96, 104, 107	generate_clusters, 19, 56, 113 GeneratedPipeline, 41, 43, 51, 54, 72, 73 GeneratedTransformation, 46, 54, 115
10,	hexagonalize, 13, 58, 101, 116
ceiling, 23, 63	
center_max, 7, 10, 16, 50, 70, 75, 76, 78, 80,	identity, 23
99, 121 center_min, 8, 9, 16, 50, 70, 75, 76, 78, 80,	is_most_centered, 11, 25, 61, 64, 67 is_most_centered(), 66
99, 121	15_most_centered(), 00
centroid, 11, 25, 62, 64, 66, 67	length(), <i>122</i>
centroid(), 3, 5, 16, 28, 30, 32, 37, 47, 49,	
85, 88, 92, 95, 103, 107	matrix multiplication, $5$
circularize, 12, 59, 101, 116	median_index, 23, 62
closest_to, 8, 10, 15, 28, 31, 33, 38, 50, 70,	median_index(), 79
75, 76, 78, 80, 99, 104, 107, 121	midrange, 11, 25, 62, 64, 67
closest_to_vec (closest_to), 15	midrange(), 3, 5, 16, 28, 30, 32, 37, 47, 49,
cluster_groups, 6, 18, 28, 33, 38, 47, 57, 83,	85, 88, 92, 95, 103, 107 min_max_scale, 65, 112
86, 89, 93, 96, 104, 107, 113	most_centered, 11, 25, 62, 64, 66
cluster_groups(), 57	most_centered(), 3, 5, 16, 28, 30, 32, 37, 47
<pre>create_dimming_fn, 21, 23, 25 create_dimming_fn(), 27</pre>	49, 61, 85, 88, 92, 95, 103, 107
create_n_fn, 22, 22, 25, 63	19, 61, 63, 66, 72, 73, 163, 167
create_n_fn(), 63, 79	pair_extremes, 8, 10, 16, 50, 68, 75, 76, 78,
create_origin_fn, 11, 22, 23, 24, 62, 64, 67	80, 99, 121
create_origin_fn(), 3, 5, 16, 28, 30, 32, 37,	<pre>pair_extremes_vec(pair_extremes), 68</pre>
47, 49, 85, 88, 92, 95, 103, 107	Pipeline, 41, 43, 51, 53, 71, 114
., ., ., ., ,,	position_max, 8, 10, 16, 50, 70, 74, 76, 78,
degrees_to_radians, 25	80, 99, 121
dim_values, 6, 16, 19, 26, 31, 33, 38, 47, 50,	position_min, 8, 10, 16, 50, 70, 75, 75, 78,
83, 86, 89, 93, 96, 104, 107	80, 99, 121
distance, 4, 16, 28, 30, 33, 38, 50, 104, 107,	
123	quantile(), $63$
dplyr::group_by(), 18	quantile_index (median_index), 62
ownerd distances 6 16 10 20 21 21 20	<pre>quantile_index(), 79</pre>
expand_distances, 6, 16, 19, 28, 31, 31, 38, 47, 50, 83, 86, 89, 93, 96, 104, 107	radians_to_degrees
	(degrees_to_radians), 25
expand_distances(), 36 expand_distances_each, 6, 16, 19, 28, 31,	rearrr::Pipeline, 42, 52
33, 36, 47, 50, 83, 86, 89, 93, 96,	rearrr::Transformation, 44, 54
104, 107	rev_windows, 8, 10, 16, 50, 70, 75, 76, 77, 80
expand_distances_each(), 32	99, 121
	<pre>rev_windows_vec (rev_windows), 77</pre>
<pre>farthest_from (furthest_from), 48</pre>	roll_elements, 8, 10, 16, 50, 70, 75, 76, 78,
FixedGroupsPipeline, 41, 44, 51, 53, 72, 73	79, 83, 99, 121
FixedGroupsTransformation, 44, 56, 115	roll_elements(), 22, 81
flip_values, 6, 19, 28, 33, 38, 46, 83, 86, 89,	<pre>roll_elements_vec (roll_elements), 79</pre>
93, 96, 104, 107	roll_values, 6, 19, 28, 33, 38, 47, 80, 81, 86
flip_values_vec (flip_values), 46	89, 93, 96, 104, 107
floor, 23, 63	roll_values(), 79
furthest_from, 8, 10, 16, 28, 31, 33, 38, 48,	roll_values_vec (roll_values), 81
70, 75, 76, 78, 80, 99, 104, 107, 121 furthest from vec (furthest from) 48	rotate_2d, 6, 19, 28, 33, 38, 47, 83, 84, 89,
TOTAL TEST TO THE VECTOR TO THE TOTAL 48	9 1 90 104 107

126 INDEX

```
rotate_3d, 6, 19, 28, 33, 38, 47, 83, 86, 87,
         93, 96, 104, 107
round, 23, 63
shear_2d, 6, 19, 28, 33, 38, 47, 83, 86, 89, 91,
         96, 104, 107
shear_3d, 6, 19, 28, 33, 38, 47, 83, 86, 89, 93,
         94, 104, 107
shuffle_hierarchy, 8, 10, 16, 50, 70, 75, 76,
         78, 80, 98, 121
square, 13, 59, 100, 116
swirl_2d, 6, 16, 19, 28, 31, 33, 38, 47, 50, 83,
         86, 89, 93, 96, 102, 107
swirl_3d, 6, 16, 19, 28, 31, 33, 38, 47, 50, 83,
         86, 89, 93, 96, 104, 105
to_unit_length, 66, 111
to_unit_length_vec(to_unit_length), 111
transfer_centroids, 19, 57, 112
Transformation, 46, 56, 114
triangularize, 13, 59, 101, 115
triplet_extremes, 8, 10, 16, 50, 70, 75, 76,
         78, 80, 99, 118
triplet_extremes_vec
         (triplet_extremes), 118
vector_length, 4, 31, 122
wrap_to_range (roll_values), 81
wrap_to_range_vec (roll_values), 81
```