

# Package ‘cocoreg’

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**Type** Package

**Title** Extract Shared Variation in Collections of Data Sets Using  
Regression Models

**Version** 0.1.1

**Description**

The algorithm extracts shared variation from a collection of data sets using regression models.

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**License** MIT + file LICENSE

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---

add_notches	<i>Add notch-like gaussian snippets to an existing signal x</i>
-------------	---

---

### Description

Add notch-like gaussian snippets to an existing signal x

### Usage

```
add_notches(x, pos, sd = 0.01 * length(x), amplitude = 1)
```

### Arguments

x	[1,N] numeric, Original data
pos	[1,m] integer, Positions to add notches to
sd	[1,1] numeric, (optional) Desired width of the Gaussian notch
amplitude	[1,1] numeric, (optional) Desired amplitude for the notches

### Value

1,N numeric, Modified signal with notches

---

apply_dc_meta	<i>Apply extracted properties of a data collection to a data collection (restore)</i>
---------------	---

---

**Description**

Apply extracted properties of a data collection to a data collection (restore)

**Usage**

```
apply_dc_meta(df_list, meta)
```

**Arguments**

df_list	list of data.frames, The data collection to process
meta,	list, Output of get_dc_meta()

**Value**

A list of data.frames, the data collection with updated metadata

---

average_R2_df1st	<i>Computes the R<sup>2</sup> (variance explained) between two lists of data.frames</i>
------------------	---

---

**Description**

Computes the R<sup>2</sup> (variance explained) between two lists of data.frames

**Usage**

```
average_R2_df1st(df_orig_lst, df_est_lst)
```

**Arguments**

df_orig_lst	List of original data.frames
df_est_lst	List of estimated data.frames

**Value**

Returns a data.frame with R2 values, one value for each data set and variable. Molten/long format.

---

 BGFA\_cocoreg\_interface

*Apply GFA using the same interface as cocoreg()*


---

### Description

Note: if K is too high GFA() might not converge in a meaningful time or the computation may mysteriously crash.

### Usage

```
BGFA_cocoreg_interface(df_list, K = 8, Nrep = 2, threshold = 0.001)
```

### Arguments

df_list	[1,m] list of data.frames, Input data to GFA in COCOREG format
K	[1,1] int, (Maximum) number of GFA components
Nrep	[1,1] int, Number of random initialization used for learning the model
threshold	[1,1] double, GFA model trimming threshold

### Value

A list with elements:

\$data:	[1,m] list of data.frames, Original data reconstructed using only those latent components that are active in all datasets
\$model:	a list, Non-trimmed output of CCAGFA::GFA()
\$dataid:	string, Dataset identifier string
\$method:	string, Analysis method identifier string
\$wall_time_taken:	[1,1] double, Time taken to run the analysis in seconds

---

 BGFA\_joint\_info

*Project BGFA components common to all datasets back to the original space*


---

### Description

Project BGFA components common to all datasets back to the original space

### Usage

```
BGFA_joint_info(model, threshold = 0.001)
```

**Arguments**

model	Output of CCAGFA::GFA()
threshold	[1,1] double, GFA model trimming threshold

**Value**

A list of data.frames, Original data reconstructed using only those latent components that are active in all datasets

---

cocoreg	<i>The Common Components by Regression (CoCoReg) algorithm</i>
---------	--

---

**Description**

An algorithm that extracts common variation between datasets using regression.

**Usage**

```
cocoreg(data, cyclic = FALSE, mapping_function = mapping_lm,
         sample_paths = FALSE, center_data = T, scale_data = F)
```

**Arguments**

data	[1,K] list of data.frames.
cyclic	boolean, Operation mode: cyclic or non-cyclic
mapping_function	function, The function to use in mappings. See mapping_lm() for an example.
sample_paths	boolean, If FALSE all paths are computed. If TRUE a subset of paths is taken: one (random) path for each starting point. Currently implemented only for cyclic=F.
center_data	boolean, Should the data be centered?
scale_data	boolean, Should the data be scaled?

**Value**

A list with elements:

\$data:	[1,K] list of data.frames containing the joint information, organised identically to the input data.
\$mappings:	[1,K*K-K] list of functions, mappings between datasets
\$paths:	[(K-1)(K-2)!, K] list of lists, paths for each data set
\$cyclic:	input cyclic as is
\$sample_paths:	boolean, TRUE if paths have been sampled, FALSE otherwise.
\$dataid:	string, Dataset identifier string
\$method:	string, Analysis method identifier string
\$wall_time_taken:	[1,1] double, Time taken to run the analysis in seconds

**Examples**

```
dc <- create_syn_data_toy()
ccr <- cocoreg(dc$data)

ggplot_dflst(dc$data, ncol=1)
ggplot_dflst(ccr$data, ncol=1)

## Not run:
ggplot_dclst(list(orig = dc$data, ccr = ccr$data))
ggplot_dclst(list(orig = dc$data, shared = ccr$data), legendMode = 'none')
ggplot_dclst(list(orig = dc$data, ccr = ccr$data), legendMode = 'all')

## End(Not run)
```

---

cocoreg_by_path	<i>Compute D_joint for dataset i separately for all paths Can be used to study path variability</i>
-----------------	---

---

**Description**

Compute D\_joint for dataset i separately for all paths Can be used to study path variability

**Usage**

```
cocoreg_by_path(data_orig, ccr, ds_ind)
```

**Arguments**

data_orig	list of data.frames, Original data collection
ccr	list, output of cocoreg(data_orig)
ds_ind	integer, index of the dataset to process

**Value**

A list of data.frames, D\_joint corresponding to each path that ends at 'ds\_ind' i.e. paths defined by ccr\$paths[,ds\_ind]. Dimensions of the matrices are the same as for the data.frames in data\_orig.

**Examples**

```
## Not run:
ccr <- cocoreg(data_list)
jibp <- cocoreg_by_path(ccr, 1)

## End(Not run)
```

---

compose	<i>Calculate the composition formed by applying all functions in the given path to a dataset.</i>
---------	---

---

**Description**

Calculate the composition formed by applying all functions in the given path to a dataset.

**Usage**

```
compose(x, path, mappings)
```

**Arguments**

x	A data frame or vector
path	A list describing the path.
mappings	A list containing the mapping functions described in the path. Usually a list of all M*M-M available mappings between the M data sets.

**Value**

A vector containing the result of the composition.

---

compose_all	<i>Calculate the average of the composition formed by applying all functions in all possible paths to a dataset.</i>
-------------	--

---

**Description**

Calculate the average of the composition formed by applying all functions in all possible paths to a dataset.

**Usage**

```
compose_all(x, paths, mappings)
```

**Arguments**

x	A list of data frames.
paths	A list of list with paths.
mappings	A list containing the mapping functions described in the path.

**Value**

A vector containing the average composition from all the paths.



---

create_mappings	<i>Generate all possible pairwise mappings between the given multivariate datasets.</i>
-----------------	---

---

### Description

The following naming convention is used in the output: '1-2' means '1' mapped to '2', i.e., '2' explained by '1'.

### Usage

```
create_mappings(data, mapping_function = mapping_lm)
```

### Arguments

data            A list of data.frames (the datasets)  
mapping\_function  
                  (optional) Default is mapping\_lm.

### Value

A named list containing the pairwise mapping functions.

---

create_syndata_mv	<i>Create multivariate synthetic data</i>
-------------------	---

---

### Description

Create multivariate synthetic data

### Usage

```
create_syndata_mv(Z, W, max_var_arr = rep(1, length(W)))
```

### Arguments

Z                [N,L] matrix, Latent factors, N observations, L factors  
W                a [1,K] list of [L,D\_k] matrices or [L,D,K] array, Projections from latent factors to data, D\_k variables per dataset  
max\_var\_arr     (optional) [1,K] numeric, Relative maximum amplitude of noise

**Value**

A list with elements:

data	Data collection as a list of data.frames
Z	Signals used
W	Mixing matrix used
E	Noise
var.coef	Noise multiplication factor used

Each dataset is a data.frame to gain compatibility with lm() and glm()

---

create\_syndata\_pwl     *A non-linear data collection using piecewise linearity*

---

**Description**

A non-linear data collection using piecewise linearity

**Usage**

```
create_syndata_pwl()
```

**Value**

A list with elements

data	Data collection as a list of data.frames
Z	Signals used
W	Mixing matrix used
Z_all	Signals shared by all datasets in the collection
Z_sub	Signals not shared by all datasets
W_all	Mixing weights for Z_all
W_sub	Mixing weights for Z_sub as a list of matrices, one matrix per dataset
E	Noise
var.coef	Noise multiplication factor used

---

create\_syn\_data\_puvar *A data collection with variables that "become unrelated during measurement"*

---

### Description

A data collection with variables that "become unrelated during measurement"

### Usage

```
create_syn_data_puvar()
```

### Value

A list with elements

data	Data collection as a list of data.frames
Z_all	Signals shared by all datasets in the collection
Z_sub	Signals not shared by all datasets
W_all	Mixing weights for Z_all
W_sub	Mixing weights for Z_sub
E	Noise
var.coef	Noise multiplication factor used

---

create\_syn\_data\_toy *An illustrative synthetic data collection*

---

### Description

Model:  $D_k = D_{\text{shared\_by\_all}} + D_{\text{shared\_by\_subset}} + D_{\text{unique}}$ ,

### Usage

```
create_syn_data_toy(N = 100, normalize = T, noisemf = 0.1)
```

### Arguments

N	Number of observations in data as integer
normalize	(optional) Should the data be processed with <code>dl_scale()</code> ? A boolean value.
noisemf	(optional) Multiplication factor for noise

**Value**

A list with elements

data	Data collection as a list of data.frames
Z_all	Signals shared by all datasets in the collection
Z_sub	Signals not shared by all datasets
W_all	Mixing weights for Z_all
W_sub	Mixing weights for Z_sub
E	Noise
var.coef	Noise multiplication factor used

**Examples**

```
## Not run:
dc <- create_syn_data_toy()
ggplot_dflst(dc$data, ncol = 1)

## End(Not run)
```

---

create\_syn\_data\_uds    *A data collection with one unrelated dataset*

---

**Description**

A data collection with one unrelated dataset

**Usage**

```
create_syn_data_uds()
```

**Value**

A list with elements

data	Data collection as a list of data.frames
Z_all	Signals shared by all datasets in the collection
Z_sub	Signals not shared by all datasets
W_all	Mixing weights for Z_all
W_sub	Mixing weights for Z_sub
E	Noise
var.coef	Noise multiplication factor used

---

create\_syn\_data\_uvar    *A collection with unrelated variables*

---

### Description

A collection with unrelated variables

### Usage

```
create_syn_data_uvar()
```

### Value

A list with elements

data	Data collection as a list of data.frames
Z_all	Signals shared by all datasets in the collection
Z_sub	Signals not shared by all datasets
W_all	Mixing weights for Z_all
W_sub	Mixing weights for Z_sub
E	Noise
var.coef	Noise multiplication factor used

---

create\_Z\_linear    *Contains functions to create synthetic datasets with different properties. The create\_syn\_data\_\*( ) functions follow the scheme: "total variation = shared\_by\_all + shared\_by\_subset + noise" Create signals*

---

### Description

Contains functions to create synthetic datasets with different properties. The create\_syn\_data\_\*( ) functions follow the scheme: "total variation = shared\_by\_all + shared\_by\_subset + noise" Create signals

### Usage

```
create_Z_linear(N, decorrelate = T)
```

### Arguments

N	Number of observations in data as integer
decorrelate	(optional) Should the variables be de-correlated?

### Value

A [N,3] matrix of signals

---

cshift	<i>Circularly shift vector elements</i>
--------	---

---

**Description**

Circularly shift vector elements

**Usage**

```
cshift(x, by)
```

**Arguments**

x	[1,N] numeric, A vector
by	[1,1] integer, How many positions to shift. by > 0 -> shift to right by = 0 -> no shift by < 0 -> shift to left

**Value**

1,N numeric, Circularly shifted signal

---

data_collections2ggdf	<i>Catenate a set of data collections (lists of data.frames) into a single melted data.frame.</i>
-----------------------	---

---

**Description**

Can be used e.g. to prepare data for plotting with ggplot().

**Usage**

```
data_collections2ggdf(..., id.vars = NULL)
```

**Arguments**

...	Several lists of data.frames to concatenate
id.vars	[1,m] string, ID variables for reshape::melt

**Value**

A data.frame with elements of ... melted and concatenated vertically into a single data.frame.

Extra columns created:

ds:	dataset id within data collection
dc:	data collection id

**Examples**

```
df_lst <- list(df1 = iris[,2:3], df2 = iris[2:3])
data_collections2ggdf(dc1 = df_lst, dc2 = df_lst)
```

---

data_matrix_rmse	<i>Compute RMSE between data.matrices dm1 and dm2</i>
------------------	---

---

**Description**

A data.matrix has observations as rows and variables as columns

**Usage**

```
data_matrix_rmse(dm1, dm2)
```

**Arguments**

dm1	[N,M] numeric, First data.matrix
dm2	[N,M] numeric, Second data.matrix

**Value**

1,M numeric, A vector of RMSE values, one per variable.

**Examples**

```
## Not run:
dm1 <- matrix(rep(1,6),nrow=2)
dm2 <- matrix(rep(3,6),nrow=2)
data_matrix_rmse(dm1, dm2)

first = list(dm1, dm1)
second = list(dm2, dm2)
(tmp = mapply(data_matrix_rmse, first, second, SIMPLIFY=FALSE))

## End(Not run)
```

---

dc_variability	<i>Compute ds_variability for all datasets in a data collection</i>
----------------	---

---

**Description**

Compute ds\_variability for all datasets in a data collection

**Usage**

```
dc_variability(data, ccr)
```

**Arguments**

data	Unprocessed dataset as a list of data.frames
ccr	Processed dataset as a list of data.frames, output of cocoreg()\$data

**Value**

Path variability as a data.frame

---

df1st2array	<i>Catenate a list of data.frames to a matrix along dim</i>
-------------	---

---

**Description**

Catenate a list of data.frames to a matrix along dim

**Usage**

```
df1st2array(df_lst, dim = 2)
```

**Arguments**

df_lst	[1,m] list of data.frames, A list of data.frames to process
dim	[1,1] int, Dimension to apply over

**Value**

A matrix with elements of df\_lst converted to matrix and catenated along dim.



---

df1st2df	<i>Catenate a list of data.frames vertically to a single data.frame</i>
----------	---

---

**Description**

Assumes equal variables for all datasets! Output has columns: <variables>, "dataset" Preserves list element names in column "dataset". For a more generic approach see df1st2dfmelt (uses reshape::melt(df\_lst))

**Usage**

```
df1st2df(df_lst, id_var_name = "dataset")
```

**Arguments**

df_lst	[1,m] list of data.frames, A list of data.frames to process
id_var_name	string, Column name for the dataset id variable

**Value**

A data.frame with elements of df\_lst catenated vertically, An extra column with dataset id is added.

---

df1st2dfmelt	<i>Combine a list of data.frames to a single molten data.frame</i>
--------------	--

---

**Description**

Output maximally "molten" with columns "dataset", "obs", "variable", "value" Preserves list element names in column "dataset".

**Usage**

```
df1st2dfmelt(df_lst)
```

**Arguments**

df_lst	[1,m] list of data.frames, A list of data.frames to process
--------	---

**Value**

A data.frame with elements of df\_lst combined using reshape::melt().

Extra columns:

**dataset** dataset name

**obs** running observation index (time)

---

df1st_add_ds	<i>Add a data.frame (dataset) to a list of data.frames (datasets)</i>
--------------	---

---

**Description**

Add a data.frame (dataset) to a list of data.frames (datasets)

**Usage**

```
df1st_add_ds(df1st, df, dsname)
```

**Arguments**

df1st	[1,m] list of data.frames, A list of data.frames
df	data.frame, Data frame to add
dsname	string, Dataset name for the data.frame to add

**Value**

A list of data.frames, A new list of data.frames with one new dataset in the end

---

df1st_dsnames2varnames	<i>Append dataset names to variable names of the respective dataset</i>
------------------------	---

---

**Description**

Append dataset names to variable names of the respective dataset

**Usage**

```
df1st_dsnames2varnames(df1st, sep = "_")
```

**Arguments**

df1st	[1,m] list of data.frames, A list of data.frames to process
sep	string, Separator to use

**Value**

A [1,m] list of data.frames with modified variable names.

---

df1st_pca	<i>Apply PCA to the data after catenating data.frames horizontally</i>
-----------	--

---

**Description**

Apply PCA to the data after catenating data.frames horizontally

**Usage**

```
df1st_pca(df_lst, center = F, scale = F)
```

**Arguments**

df_lst	[1,m] list of data.frames, A list of data.frames to process
center	boolean, TRUE -> center data, FALSE -> do nothing
scale	boolean, TRUE -> scale data, FALSE -> do nothing

**Value**

A list with elements:

pcdf:	data.frame, PCA components (prcomp()\$x)
model:	list, Output of prcomp()

---

df_ggplot_melt	<i>Melt data.frame into ggplottable format</i>
----------------	--

---

**Description**

Melts a data.frame into format that is suitable for use with ggplot2. Creates the time variable 't' used by plotting functions.

**Usage**

```
df_ggplot_melt(df)
```

**Arguments**

df	A data.frame
----	--------------

**Value**

A ggplot2 compatible data.frame with time variable

**Examples**

```
## Not run:
dc <- create_syn_data_toy()
df <- dc$data[[1]]
str(df)
str(df_ggplot_melt(df))

## End(Not run)
```

---

**df\_scale***Apply scale on a numeric data.frame*

---

**Description**

Apply scale on a numeric data.frame

**Usage**

```
df_scale(df, ...)
```

**Arguments**

df                    data.frame, A numeric data.frame to process  
...                    arguments to scale()

**Value**

data.frame, A scaled data.frame with attributes preserved

---

**df\_scale\_ols***Scales variables in data.frame dfx using ordinary least squares such*

---

**Description**

Scales variables in data.frame dfx using ordinary least squares such that the scaled result explains as much of the variance in dfy as possible. Scaling is done separately for each variable (i.e. no linear mixing of variables). Assumes data.frames dfx and dfy to be of identical structure. Intended use: to scale up cocoreg projections to account for the lost variance.

**Usage**

```
df_scale_ols(dfx, dfy)
```

**Arguments**

dfx                    data.frame, Data frame to use as independent variable  
dfy                    data.frame, Data frames to use as dependent variable

**Value**

data.frame, A rescaled version of dfx with dimnames from dfy.

**Examples**

```
## Not run:  
dc <- create_syn_data_toy()  
ccr <- cocoreg(dc$data)  
dfLst <- mapply(df_scale_ols, ccr$data, dc$data , SIMPLIFY=F)  
  
## End(Not run)
```

---

dl\_remove\_NA                    *Remove rows with NA values from a list of data.frames*

---

**Description**

Same rows removed from all data frames in the list.

**Usage**

```
dl_remove_NA(df_lst)
```

**Arguments**

df\_lst                    [1,m] list of data.frames, A list of data.frames to process

**Value**

1,m list of data.frames, Data without rows that contain NA, same rows removed from all data.frames in the input list.

---

dl_scale	<i>Run scale() on a list of data.frames</i>
----------	---

---

**Description**

Run scale() on a list of data.frames

**Usage**

```
dl_scale(dl, ...)
```

**Arguments**

dl	A list of data.frames
...	Additional arguments to be passed on to scale

**Value**

A list of data.frames that have been processed using scale() and converted back to data.frame

---

ds_variability	<i>Compute variability_components for a dataset</i>
----------------	---

---

**Description**

Note: might not work for cyclic ccr <TODO>

**Usage**

```
ds_variability(data, ccr, ds_ind)
```

**Arguments**

data	Unprocessed dataset as a list of data.frames
ccr	Processed dataset as a list of data.frames, output of cocoreg()\$data
ds_ind	Starting dataset of the set of paths to analyze as [1,1] integer

**Value**

Path variability as a data.frame

---

`generate_mapping_function`*Generate a mapping function between two datasets*

---

**Description**

Generate a mapping function between two datasets, using some method, such as linear regression (lm), or some classifier such as a random forest (randomForest).

Wraps the function as well as data into a single object.

**Usage**

```
generate_mapping_function(method = lm)
```

**Arguments**

`method` A function to be used in the mapping. A function object.

**Value**

Returns a function object that does the mapping between two datasets i.e. from dataset 1 to dataset 2.

---

`generate_paths`*Generate all/some paths between points*

---

**Description**

Generate all/some paths between points

**Usage**

```
generate_paths(ind, n, cyclic = FALSE, sample_paths = FALSE)
```

**Arguments**

`ind` [1,2] int, The starting and ending point c(start, end).

`n` [1,1] int, Number of points in the whole set.

`cyclic` boolean, Should the path be cyclic (1-2-1) or noncyclic (2-1).

`sample_paths` boolean, If FALSE, all possible paths are generated. If true one path per ending point is selected.

**Value**

A list of lists containing the paths.

---

generate\_paths\_cyclic *Generate cyclic paths*

---

**Description**

From a set of n numbers, generate all possible paths starting from and ending on a given number.

**Usage**

```
generate_paths_cyclic(ind, n)
```

**Arguments**

ind	[1,1] ind, The starting dataset (equals to ending dataset because of cycle).
n	[1,1] ind, The number of datasets.

**Value**

A list of lists containing the paths.

---

generate\_paths\_noncyclic  
*Generate non-cyclic paths*

---

**Description**

From a set of n numbers, generate all possible paths starting from and ending on a given number.

**Usage**

```
generate_paths_noncyclic(ind, n, sample_paths = FALSE)
```

**Arguments**

ind	The starting dataset
n	The number of datasets.
sample_paths	boolean, If FALSE, all possible paths are generated. If TRUE one path per ending point is selected.

**Value**

A list of lists containing the paths.



---

get_dc_meta	<i>Extract important properties of data collection</i>
-------------	--

---

**Description**

Extract important properties of data collection

**Usage**

```
get_dc_meta(df_list, type = "current")
```

**Arguments**

df_list	list of data.frames, The data collection to process
type	string, If 'current' then data collection metadata is collected from the data collection itself. If 'original' then metadata is collected from special attributes.

**Value**

A list with elements:

\$dcnames:	Dataset names
\$dcdimnames:	A list of dataset dimension names for each dataset

---

get_range_datalist	<i>Get [min, max] of a list of numeric objects</i>
--------------------	--

---

**Description**

Get [min, max] of a list of numeric objects

**Usage**

```
get_range_datalist(dataList)
```

**Arguments**

dataList	[1,m] list of numeric objects
----------	-------------------------------

**Value**

1,2 double, [min, max] of the input

---

`get_starting_dataset` *Helper function to get the starting dataset based on a path.*

---

**Description**

Helper function to get the starting dataset based on a path.

**Usage**

```
get_starting_dataset(p)
```

**Arguments**

`p` A path.

**Value**

A number indicating the starting dataset

---

`ggcompare_dclst` *Compare data collections variable by variable*

---

**Description**

Compare data collections variable by variable

**Usage**

```
ggcompare_dclst(dclst)
```

**Arguments**

`dclst` A (named) list of data collections i.e. a list of lists of data.frames (see examples)

**Value**

Returns a ggplot object (which is by default printed if not assigned to variable)

---

**ggplot\_dclst** *Plotting data collections using ggplot*

---

**Description**

Plotting data collections using ggplot

**Usage**

```
ggplot_dclst(dc_lst, ylim = NULL, titleArr = names(dc_lst),  
            legendMode = "none", dfplot = ggplot_df)
```

**Arguments**

dc_lst	A list of data collections i.e. a list of lists of data.frames (see examples)
ylim	(optional) y-axis limits as [1,2] numeric, passed on to dfplot() as 'ylim'
titleArr	(optional) Plot column titles as [1, length(dc_lst)] string array
legendMode	(optional) Where to put legend, allowed values c('none', 'first', 'all')
dfplot	(optional) Function used to plot a data.frame (one panel in final plot)

**Value**

Produces a plot to the active graphics device

**Examples**

```
## Not run:  
dc <- create_syn_data_toy()  
ccr <- cocoreg(dc$data)  
ggplot_dclst(list(d1 = dc$data, d2 = ccr$data, dn = dc$data))  
  
## End(Not run)
```

---

**ggplot\_df** *Plotting data.frame using ggplot*

---

**Description**

Plotting data.frame using ggplot

**Usage**

```
ggplot_df(df, titlestr = NULL, ylabstr = NULL, ylim = NULL,  
          color = NULL, linetype = NULL, logy = F)
```

**Arguments**

df	A data.frame to plot
titlestr	(optional) Title of plot as string
ylabstr	(optional) Y-axis label as string
ylim	(optional) y-axis limits as [1,2] numeric, passed on to dfplot() as 'ylim'
color	(optional) Input for manual color scale
linetype	(optional) Input for manual linetype scale
logy	(optional) Should y-axis be logarithmic? A boolean value.

**Value**

Returns a ggplot2 object

**Examples**

```
## Not run:
dc <- create_syn_data_toy()
ggplot_df(dc$data[[1]])

## End(Not run)
```

---

ggplot\_dflst

*Plot a list of data.frames using ggplot2*


---

**Description**

Plot a list of data.frames using ggplot2

**Usage**

```
ggplot_dflst(dflst, ncol = 1, plot = T, plotfun = ggplot_df, ...)
```

**Arguments**

dflst	A list of datasets as a list of data.frames
ncol	(optional) Number of columns in final plot
plot	(optional) Plot or not: if TRUE produces a plot else returns a list of ggplot objects
plotfun	(optional) Function used to plot a data.frame (one panel in final plot)
...	(optional) Additional parameters passed on to plotfun

**Value**

Produces a plot to the active graphics device or returns a list of ggplot objects

**Examples**

```
## Not run:  
dc <- create_syn_data_toy()  
ggplot_dflst(dc$data)  
  
## End(Not run)
```

---

make\_data\_gauss\_2d      *Make 2D gauss data (maybe obsolete)*

---

**Description**

Make 2D gauss data (maybe obsolete)

**Usage**

```
make_data_gauss_2d(n, var, angle_deg, scale = T, seed = 42)
```

**Arguments**

n	[1,1] int, Number of observations
var	[1,2] numeric, Variances
angle_deg	[1,1] numeric, Rotation angle
scale	boolean, Scale data? T -> scale, F -> do not scale
seed	[1,1] int, Random seed

**Value**

Matrix of 2D gaussian data

---

mappings\_R2\_matrix      *Extract R2 values from a list of mappings using summary()*

---

**Description**

Extract R2 values from a list of mappings using summary()

**Usage**

```
mappings_R2_matrix(mappings, n_datasets, aggfun = mean)
```

**Arguments**

mappings	[M*M-M] list, Exhaustive list of mappings between the M datasets
n_datasets	integer, Number of datasets i.e. M
aggfun	function, A function to apply when aggregating R2 values over variables in a multivariate dataset

**Value**

A [M,M] matrix of R2 values stored such that the R2 value for mapping a->b is read from row a and column b.

**Examples**

```
## Not run:  
ccr <- cocoreg(data_collection)  
R2mat <- mappings_R2_matrix(ccr$mappings, length(ccr$data))  
  
## End(Not run)
```

---

mapping_glmnet	<i>Define a mapping function using glmnet::glmnet</i>
----------------	---

---

**Description**

Define a mapping function using glmnet::glmnet

**Usage**

```
mapping_glmnet(data1, data2)
```

**Arguments**

data1	data.frame, Dataset 1, the independent variables
data2	data.frame, Dataset 2, the dependent variables

**Value**

Returns a function object that does the mapping between two datasets.

---

mapping_lm	<i>Mapping stats::lm</i>
------------	--------------------------

---

**Description**

Mapping stats::lm

**Usage**

```
mapping_lm(data1, data2)
```

**Arguments**

data1	data.frame, Dataset 1, the independent variables
data2	data.frame, Dataset 2, the dependent variables

---

mapping_lmridge	<i>Define a mapping function using MASS::lm.ridge</i>
-----------------	---

---

**Description**

Define a mapping function using MASS::lm.ridge

**Usage**

```
mapping_lmridge(data1, data2)
```

**Arguments**

data1	data.frame, Dataset 1, the independent variables
data2	data.frame, Dataset 2, the dependent variables

**Value**

Returns a function object that does the mapping between two datasets.

---

`mapping_pcr`*Define a mapping function using pls::pcr*

---

**Description**

Define a mapping function using pls::pcr

**Usage**

```
mapping_pcr(data1, data2)
```

**Arguments**

<code>data1</code>	data.frame, Dataset 1, the independent variables
<code>data2</code>	data.frame, Dataset 2, the dependent variables

**Value**

Returns a function object that does the mapping between two datasets.

---

`mapping_rf`*Mapping randomForest*

---

**Description**

Mapping randomForest

**Usage**

```
mapping_rf(data1, data2)
```

**Arguments**

<code>data1</code>	data.frame, Dataset 1, the independent variables
<code>data2</code>	data.frame, Dataset 2, the dependent variables



---

mapping_rlm	<i>Mapping MASS::rlm</i>
-------------	--------------------------

---

**Description**

Mapping MASS::rlm

**Usage**

```
mapping_rlm(data1, data2)
```

**Arguments**

data1	data.frame, Dataset 1, the independent variables
data2	data.frame, Dataset 2, the dependent variables

---

---

mapping_svm	<i>Mapping svm</i>
-------------	--------------------

---

**Description**

Mapping svm

**Usage**

```
mapping_svm(data1, data2)
```

**Arguments**

data1	data.frame, Dataset 1, the independent variables
data2	data.frame, Dataset 2, the dependent variables

---

---

mapping_svm_sigmoid	<i>Mapping svm using sigmoid</i>
---------------------	----------------------------------

---

**Description**

Mapping svm using sigmoid

**Usage**

```
mapping_svm_sigmoid(data1, data2)
```

**Arguments**

data1	data.frame, Dataset 1, the independent variables
data2	data.frame, Dataset 2, the dependent variables

---

<code>matrix_variability</code>	<i>Compute "variance" of the matrices using Frobenius norm. Variance is by default computed with respect to the mean of the matrices.</i>
---------------------------------	---

---

**Description**

Compute "variance" of the matrices using Frobenius norm. Variance is by default computed with respect to the mean of the matrices.

**Usage**

```
matrix_variability(mat_lst, mean_mat = apply(abind::abind(mat_lst, along = 3),
  c(1, 2), mean))
```

**Arguments**

<code>mat_lst</code>	A [1,M] list of [I,J] matrices from which variability should be computed
<code>mean_mat</code>	A [I,J] matrix describing the mean observation for <code>mat_lst</code> .

**Value**

A list with elements

<code>fbnorm</code>	Frobenius norm values for each of the matrices
<code>diff</code>	[I,J,M] matrix of differences <code>mat_lst-mean_mat</code>

---

<code>nplst_reorder_grid</code>	<i>Reorders a nested list of ggplots</i>
---------------------------------	--

---

**Description**

Reorders a nested list of ggplots to `ncol` columns prior to calling `grid.arrange()` Note: `p_list` is a list of lists of ggplots. `p_list = list(p_list1, p_list2,...)`

**Usage**

```
nplst_reorder_grid(p_list, ncol)
```

**Arguments**

<code>p_list</code>	A list of lists of ggplots
<code>ncol</code>	Target number of columns, integer value

**Value**

A reordered and flattened version of input list as a list of ggplot2 objects

---

PCA\_cocoreg\_interface *PCA projection using cocoreg interface*

---

**Description**

PCA projection using cocoreg interface

**Usage**

```
PCA_cocoreg_interface(df_list, prc_th = 0.9)
```

**Arguments**

df_list	[1,m] list of data.frames, Input data to GFA in COCOREG format
prc_th	[1,1] double, Threshold in percentage of cumulative variance explained PCA components are included until cumulative explained variance reaches prc_th.

**Value**

A list with elements:

\$data	[1,m] list of data.frames, Original data reconstructed using only those latent components that are active in all datasets
\$dataid	string, Dataset identifier string
\$method	string, Analysis method identifier string
\$wall_time_taken	[1,1] double, Time taken to run the analysis in seconds

---

rename\_variables *Rename variables of a data collection*

---

**Description**

Rename variable in all datasets such that the data.frame list conforms to the requirements of Co-CoReg.

**Usage**

```
rename_variables(df_list)
```

**Arguments**

df_list	list of data.frames, The datasets to process
---------	--

**Value**

A list of data.frames with changed variable names. Original dimension names are stored as an attribute.

---

repmat	<i>Replicate matrix to create a larger one</i>
--------	--

---

**Description**

From: <http://haky-functions.blogspot.fi/2006/11/repmat-function-matlab.html> (accessed 27.3.2015)

**Usage**

```
repmat(X, m, n)
```

**Arguments**

X	A [I,J] matrix or J element vector, Matrix used as such, vector coerced to a row matrix with $\text{dim}(X)=[1,J]$ .
m	[1,1] integer, Replication count vertically
n	[1,1] integer, Replication count horizontally

**Value**

$m \times I, n \times J$  matrix, Replicated data

---

RGCCA\_cocoreg\_interface

*COCOREG style analysis using RGCCA projection*

---

**Description**

COCOREG interface used for both input and output.

**Usage**

```
RGCCA_cocoreg_interface(df1st, tauArr = rep(0.5, length(df1st)))
```

**Arguments**

df1st,	[1,m] list of data.frames, Input data to GFA in COCOREG format
tauArr,	[1,m] double, See RGCCA::rgcca() for details

**Value**

A list with elements:

\$data	[1,m] list of data.frames, Original data reconstructed using only those latent components that are active in all datasets
\$model	list, The output RGCCA::rgcca()
\$dataid	string, Dataset identifier string
\$method	string, Analysis method identifier string
\$wall_time_taken	[1,1] double, Time taken to run the analysis in seconds

---

rmse	<i>Compute RMSE between vectors v1 and v2</i>
------	---

---

**Description**

Compute RMSE between vectors v1 and v2

**Usage**

```
rmse(v1, v2, relative = F)
```

**Arguments**

v1	[1,m] numeric, First data vector
v2	[1,m] numeric, Second data vector
relative	boolean, If TRUE, relate the rmse value to the rmse of v1. If FALSE, just compute RMSE between v1 and v2

**Value**

1,1 double, RMSE value

---

rotation_matrix	<i>A rotation matrix</i>
-----------------	--------------------------

---

**Description**

A rotation matrix

**Usage**

```
rotation_matrix(angle_deg)
```

**Arguments**

angle\_deg [1,1] numeric, Angle in degrees

**Value**

2,2 matrix, Rotation matrix for making angle\_deg 2D rotation

---

row_suffle_variability	<i>Determine the variability of matrices under row shuffling</i>
------------------------	--

---

**Description**

Determine the variability of matrices under row shuffling

**Usage**

```
row_suffle_variability(mat_lst, B = 50)
```

**Arguments**

mat\_lst A list of matrices  
B integer, Number of times to sample (shuffle)

**Value**

B,K matrix, Frobenius norm vectors corresponding to row shuffling

---

SCA\_cocoreg\_interface *SCA projection using cocoreg interface*

---

**Description**

SCA projection using cocoreg interface

**Usage**

```
SCA_cocoreg_interface(df_list, nfac = 1, type = "sca-p",
  rotation = "none", nstart = 10)
```

**Arguments**

df_list,	[1,m] list of data.frames, Input data to GFA in COCOREG format
nfac,	[1,1] int, see multiway::sca() for details
type,	string, Type of analysis, see multiway::sca() for details
rotation,	string, see multiway::sca() for details
nstart,	[1,1] int, see multiway::sca() for details

**Value**

A list with elements:

\$data	[1,m] list of data.frames, Original data reconstructed using only those latent components that are active in all datasets
\$model	list, The output of multiway::sca()
\$dataid	string, Dataset identifier string
\$method	string, Analysis method identifier string
\$wall_time_taken	[1,1] double, Time taken to run the analysis in seconds

---

se	<i>Standard error of mean</i>
----	-------------------------------

---

**Description**

Standard error of mean

**Usage**

```
se(x, na.rm = T)
```

**Arguments**

x [1,M] numeric, data vector  
 na.rm procedure for NA's, passed on to sd(), default: na.rm = T

**Value**

1,1 numeric, standard error of mean

---

ss	<i>Sum of squares</i>
----	-----------------------

---

**Description**

Sum of squares

**Usage**

ss(x)

**Arguments**

x [1,m] numeric, A data vector

**Value**

Sum of squares of x

---

svm_sigmoid	<i>SVM using sigmoid kernel</i>
-------------	---------------------------------

---

**Description**

SVM using sigmoid kernel

**Usage**

svm\_sigmoid(...)

**Arguments**

... Further arguments passed on to e1071::svm()



---

to_unit_vec	<i>Make vector of unit norm</i>
-------------	---------------------------------

---

**Description**

Make vector of unit norm

**Usage**

```
to_unit_vec(x)
```

**Arguments**

x	[1,m] double, A vector of data
---	--------------------------------

**Value**

1,m double, Same vector normalized to unit Euclidean norm

---

traverse_nested_list	<i>Apply fun to the bottom level of a nested list structure</i>
----------------------	---

---

**Description**

Used to batch process computation results that are stored into a nested list structure. Analysis results are stored as lists but with class attribute changed. This signals that the recursion into the list structure should end and fun should be applied instead. Can be used e.g. to pick out results from a complex list structure.

**Usage**

```
traverse_nested_list(lst, fun, exclude_names = NULL, ...)
```

**Arguments**

lst	nested list, A nested list structure to process
fun	function object, The function to apply at the bottom level
exclude_names	string array, Names of list elements to skip at any level
...	Further parameters passed on to fun

**Value**

A list outputs generated when applying fun to the bottom level of input lst. Bottom level is considered reached when something other than class == 'list' is encountered.

---

validate_data	<i>Validate a data collection for use with cocoreg</i>
---------------	--

---

**Description**

Check that the data collection has all the required properties.

**Usage**

```
validate_data(df_list)
```

**Arguments**

df\_list            list of data.frames, The data collection to validate

**Value**

A list of data.frames that conform to the requirements

---

variability_components	<i>Compute total, within group and between group variability using fun</i>
------------------------	--

---

**Description**

The function used the definition:  $gvar = tvar - wgvar$

**Usage**

```
variability_components(vec, grp, fun)
```

**Arguments**

vec                [1,M] numeric, Data vector  
 grp               [1,M] integer/character vector, Some grouping of vec  
 fun                function, Function to use when quantifying the variability

**Value**

A list with elements:

tvar:              Total variability  
 bgvar:            Between groups variability,  $tvar - \text{sum}(wgvar\_*)$   
 wgvar\_<groupname>:    Within group variability for each group  
 wg\_rel:             $\text{sum}(wgvar)/tvar$   
 bg\_rel:             $bgvar/tvar$

**Examples**

```
vec <- rnorm(10)
grp <- rep(c("a","b","c"), c(3,3,4))
variability_components(vec, grp, ss)
```

---

variation\_shared\_by *Return a specific variation component*

---

**Description**

Variation can be shared by: 'all' all datasets 'subset' a subset of the datasets (excluding variation already in 'all') 'all\_and\_subset' a union of the above

The returned data never contains noise (which is considered to be part of each datasets unique variation). The linear toy datasets do not contain variation unique to a dataset other than pure noise.

**Usage**

```
variation_shared_by(dc, type, center = T, scale = F)
```

**Arguments**

dc	A data collection from one of the create_syn_data_*() functions
type	Type of variation to extract, allowed values c('all','subset','all_and_subset')
center	(optional) Should the output data be centered?
scale	(optional) Should the output data be scaled?

**Value**

A list of data.frames containing the desired variation component

**Examples**

```
## Not run:
dc <- create_syn_data_toy()
ldSharedByAll = variation_shared_by(dc, "all", center = F)
ldSharedBySome = variation_shared_by(dc, "subset", center = F)
ldNonUnique = variation_shared_by(dc, "all_and_subset", center = F)
dNoise <- mapply(function(x,y){x-y}, x=dc$data, y=ldNonUnique, SIMPLIFY = F)
ggplot_dclst(list(observed = dc$data,
                 shared.by.all = ldSharedByAll,
                 shared.by.some = ldSharedBySome,
                 noise = dNoise),
            ylim = c(-3, 3))

## End(Not run)
```

---

var_explained	<i>Sum-of-squares values showing what portion of variance in dvec is explained by dvec_est</i>
---------------	--

---

### Description

Computation as in: [http://en.wikipedia.org/wiki/Fraction\\_of\\_variance\\_unexplained](http://en.wikipedia.org/wiki/Fraction_of_variance_unexplained)

ss\_est becomes zero if dvec\_est equals dvec\_0= $\text{rep}(\text{mean}(\text{dvec}), \text{length}(\text{dvec}))$ . If dvec\_est is better estimate than dvec\_0, R2 is positive. If dvec\_est is worse than dvec\_0, R2 is negative.

### Usage

```
var_explained(dvec, dvec_est)
```

### Arguments

dvec	[1,m] numeric, data vector
dvec_est	[1,m] numeric, data vector, an estimate of dvec

### Value

A list with elements:

ss_tot:	Sum of squares in dvec
ss_est:	Sum of squares in dvec_est
ss_err:	Sum of squares of dvec - dvec_est
R2:	Percentage of variance explained i.e. $1 - \text{ss\_err}/\text{ss\_tot}$

---

vecnorm	<i>Compute Euclidean norm of vector</i>
---------	---

---

### Description

Convenience function for use with e.g. lapply

### Usage

```
vecnorm(x)
```

### Arguments

x	[1,m] double, A vector of data
---	--------------------------------

### Value

1,1 double, Euclidean norm of x

---

vector\_variability      *Compute "variance" of the vectors var()*

---

**Description**

Compute "variance" of the vectors var()

**Usage**

```
vector_variability(vec_lst, mean_vec = apply(abind::abind(vec_lst, along = 2),
  1, mean))
```

**Arguments**

vec\_lst                  Data to process as a list of numeric vectors  
 mean\_vec                (optional) Desired mean vector as a numeric vector

**Value**

Variance of data values around mean as a numeric matrix

---

wrapper\_BGFA              *Run BGFA by Klami et. al. using data format conventions of this repo*

---

**Description**

Run BGFA by Klami et. al. using data format conventions of this repo

**Usage**

```
wrapper_BGFA(df_list, K = 8, Nrep = 2)
```

**Arguments**

df\_list                  [1,m] list of data.frames, Input data to GFA in COCOREG format  
 K                        [1,1] int, The (maximum) number of components; should be high enough to capture all of the components. This can be recognized by at least a few of the components being shut down  
 Nrep                    [1,1] int, Number of random initialization used for learning the model

**Value**

A list, The output of CCAGFA::GFA()

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