Package ‘xnet’

February 3, 2020

Type Package
Title Two-Step Kernel Ridge Regression for Network Predictions
Version 0.1.11
Date 2020-02-03
BugReports https://github.com/CenterForStatistics-UGent/xnet/issues
URL https://github.com/CenterForStatistics-UGent/xnet
Depends R(>= 3.4.0)
Imports methods, utils, graphics, stats, grDevices
License GPL-3
Encoding UTF-8
LazyData true
RoxygenNote 7.0.2
Suggests testthat, knitr, rmarkdown, ChemmineR, covr, fmcsR
VignetteBuilder knitr
Collate 'Class_linearFilter.R' 'all_generics.R' 'Class_permtest.R'
   'Class_tskrr.R' 'Class_tskrrHeterogeneous.R'
   'Class_tskrrHomogeneous.R' 'Class_tskrrImpute.R'
   'Class_tskrrImputeHeterogeneous.R'
   'Class_tskrrImputeHomogeneous.R' 'Class_tskrrTune.R'
   'Class_tskrrTuneHeterogeneous.R' 'Class_tskrrTuneHomogeneous.R'
   'as_tuned.R' 'create_grid.R' 'data_drugtarget.R'
   'data_proteinInteraction.R' 'dim.R' 'eigen2hat.R' 'fitted.R'
   'get_loo_fun.R' 'getlooInternal.R' 'getters_linearFilter.R'
   'getters_permtest.R' 'getters_tskrr.R' 'getters_tskrrImpute.R'
   'getters_tskrrTune.R' 'hat.R' 'impute_tskrr.R'
   'impute_tskrr.fit.R' 'internal_helpers.R' 'is_symmetric.R'
   'labels.R' 'linear_filter.R' 'loo.R' 'looInternal.R' 'loss.R'
R topics documented:

- 'loss_functions.R'
- 'match_labels.R'
- 'permtest.R'
- 'plot.tskrr.R'
- 'plot_grid.R'
- 'predict.R'
- 'prepare_lambdas.R'
- 'residuals.R'
- 'test_input.R'
- 'test_symmetry.R'
- 'tskrr.R'
- 'tskrr.fit.R'
- 'tune.R'
- 'update.R'
- 'valid_dimensions.R'
- 'valid_labels.R'
- 'weights.R'
- 'xnet-package.R'

**NeedsCompilation** no

**Author** Joris Meys [cre, aut],
Michiel Stock [aut]

**Maintainer** Joris Meys <Joris.Meys@UGent.be>

**Repository** CRAN

**Date/Publication** 2020-02-03 19:30:02 UTC

---

R topics documented:

- xnet-package ................................................. 3
- alpha ......................................................... 4
- as_tuned ...................................................... 5
- create_grid .................................................. 6
- dim,tskrr-method ............................................ 7
- drugTargetInteraction ......................................... 7
- eigen2hat ..................................................... 8
- fitted.tskrr .................................................. 9
- get_loo_fun ................................................... 10
- has_imputed_values ........................................... 12
- hat .......................................................... 13
- impute_tskrr ................................................. 13
- impute_tskrr.fit .............................................. 15
- is_symmetric .................................................. 16
- is_tuned ..................................................... 17
- labels.tskrr ................................................. 18
- linearFilter-class ........................................... 19
- linear_filter ................................................. 20
- loo .......................................................... 21
- loo_internal .................................................. 22
- loss .......................................................... 23
- loss_functions ............................................... 24
- match_labels .................................................. 25
- permtest ...................................................... 26
- permtest-class ................................................. 27
- permutations ................................................... 28
- plot.tskrr ..................................................... 29
- plot_grid ...................................................... 30
- predict.tskrr .................................................. 32
- proteinInteraction ............................................ 36
- residuals ...................................................... 37
- response.tskrr ................................................ 38
Description

This package implements the two-step kernel ridge regression model, a supervised network prediction method that can be used for all kinds of network analyses. Examples are protein-protein interaction, foodwebs, ...

Author(s)

Joris Meys and Michiel Stock

See Also

Send your bug reports to:

https://github.com/CenterForStatistics-UGent/xnet/issues

More background in the paper by Stock et al, 2018:

http://doi.org/10.1093/bib/bby095
Getters for linearFilter objects

Description

These functions allow you to extract slots from objects of the class `linearFilter`.

Usage

```r
alpha(x)

na_removed(x)

## S3 method for class 'linearFilter'
mean(x, ...)

## S4 method for signature 'linearFilter'
mean(x, ...)

## S4 method for signature 'linearFilter'
colMeans(x)

## S4 method for signature 'linearFilter'
rowMeans(x)

## S4 method for signature 'linearFilter'
alpha(x)

## S4 method for signature 'linearFilter'
na_removed(x)
```

Arguments

- `x`: a `linearFilter` object
- `...`: arguments passed to or from other methods.

Value

- for `mean`: the mean of the original matrix
- for `colMeans`: a numeric vector with the column means
- for `rowMeans`: a numeric vector with the row means
- for `alpha`: a numeric vector of length 4 with the alpha values.
- for `na_removed`: a logical value indicating whether missing values were removed prior to the fitting of the filter.
Examples

data(drugTarget)
lf <- linear_filter(drugTargetInteraction, alpha = 0.25)
alpha(lf)
mean(lf)
colMeans(lf)
na_removed(lf)

Description

These functions allow converting models that inherit from the `tskrr` and `tskrrTune` class into each other, keeping track of whether the model is homogeneous or heterogeneous. The dots argument allows specifying values for possible extra slots when converting from `tskrr` to `tskrrTune`. More information on these slots can be found on the help page of `tskrrTune`. These functions are not exported.

Usage

as_tuned(x, ...)
as_tskrr(x, ...)

## S4 method for signature 'tskrrHomogeneous'
as_tuned(x, ...)

## S4 method for signature 'tskrrHeterogeneous'
as_tuned(x, ...)

## S4 method for signature 'tskrrTune'
as_tskrr(x)

## S4 method for signature 'tskrrImpute'
as_tskrr(x)

## S4 method for signature 'tskrr'
as_tskrr(x)

Arguments

x           a model of class `tskrr`
...
    values for the extra slots defined by the class `tskrrTune`
create_grid

Value
For as_tuned: a tskrrTune object of the proper class (homogeneous or heterogeneous)
For as_tskrr: an object of class tskrrHomogeneous or tskrrHeterogeneous depending on whether
the original object was homogeneous or heterogeneous.

Warning
This functions do NOT tune a model. they are used internally to make the connection between both
types in the methods.

See Also
• tune for actually tuning a model.
• tskrrTune for names and possible values of the slots passed through …

create_grid

Create a grid of values for tuning tskrr

Description
This function creates a grid of values for tuning a tskrr model. The grid is equally spaced on a
logarithmic scale. Normally it’s not needed to call this method directly, it’s usually called from
tune.

Usage
create_grid(lim = c(1e-04, 10000), ngrid = 10)

Arguments
lim a numeric vector with 2 values giving the lower and upper limit for the grid.
ngrid the number of values that have to be produced. If this number is not integer, it is
truncated. The value should be 2 or larger.

Details
The lim argument sets the boundaries of the domain in which the lambdas are sought. The lambda
values at which the function is evaluated, are calculated as:
exp(seq(log(1e-4),log(1e4),length.out = ngrid))

Value
a numeric vector with values evenly spaced on a logarithmic scale.

See Also
tune for tuning a tskrr model.
**Examples**

```r
create_grid(lim = c(1e-4, 1), ngrid = 5)
```

---

**dim, tskrr-method**  
*Get the dimensions of a tskrr object*

**Description**

These functions allow you to extract the dimensions of a tskrr object. These dimensions are essentially the dimensions of the label matrix y.

**Usage**

```r
## S4 method for signature 'tskrr'

dim(x)
```

**Arguments**

- `x`  
a tskrr object.

**Value**

a vector with two values indicating the number of rows and the number of columns.

**Examples**

```r
data(drugtarget)
mod <- tskrr(drugTargetInteraction, targetSim, drugSim)

dim(mod)
nrow(mod)
ncol(mod)
```

---

**drugTargetInteraction**  
*drug target interactions for neural receptors*

**Description**

A dataset for examining the interaction between 54 drugs and 26 neural receptors. It consists of three different matrices.

**Usage**

```r
drugTargetInteraction
```
Format

- for drugTargetInteraction: a numeric matrix of 26 rows by 54 columns.
- For drugSim: a numeric square matrix with 54 rows/columns.
- For targetSim: a numeric square matrix with 26 rows/columns.

Details

The dataset consists of the following objects:

- drugTargetInteraction: a matrix indicating whether or not a certain drug compound interacts with a certain neural receptor.
- targetSim: a similarity matrix for the neural receptors.
- drugSim: a similarity matrix for the drugs

The data originates from Yamanishi et al (2008) but was partly reworked to be suitable for two-step kernel ridge regression. This is explained in detail in the Preparation of the example data vignette.

Source

https://doi.org/10.1093/bioinformatics/btn162

References


eigen2hat

Calculate the hat matrix from an eigen decomposition

description

These functions calculate either the hat matrix, the mapping matrix or the original (kernel) matrix for a two-step kernel ridge regression, based on the eigendecomposition of the kernel matrix.

Usage

eigen2hat(eigen, val, lambda)

eigen2map(eigen, val, lambda)

eigen2matrix(eigen, val)

Arguments

eigen a matrix with the eigenvectors.
val an numeric vector with the eigenvalues.
lambda a single numeric value for the hyperparameter lambda
Details

For the hat matrix, this boils down to:

\[ U \Sigma (\Sigma + \lambda I)^{-1} U^T \]

For the map matrix, this is:

\[ U (\Sigma + \lambda I)^{-1} U^T \]

with \( U \) the matrix with eigenvectors, \( \Sigma \) a diagonal matrix with the eigenvalues on the diagonal, \( I \) the identity matrix and \( \lambda \) the hyperparameter linked to this kernel. The internal calculation is optimized to avoid having to invert a matrix. This is done using the fact that \( \Sigma \) is a diagonal matrix.

Value

a numeric matrix representing either the hat matrix (eigen\text{hat}), the map matrix (eigen\text{map}) or the original matrix (eigen\text{matrix})

---

fitted.tskrr  extract the predictions

Description

This function extracts the fitted predictions from a tskrr object or an object inheriting from that class. The xnet package provides an S4 generic for the function fitted from the package stats, and a method for tskrr objects.

Usage

```r
## S3 method for class 'tskrr'
fitted(object, labels = TRUE, ...)
## S3 method for class 'linearFilter'
fitted(object, ...)
## S4 method for signature 'tskrr'
fitted(object, labels = TRUE, ...)
## S4 method for signature 'linearFilter'
fitted(object, ...)
```

Arguments

- **object**
  - an object for which the extraction of model fitted values is meaningful.
- **labels**
  - a logical value indicating whether the labels should be shown. Defaults to TRUE
- **...**
  - arguments passed to or from other methods.
get_loo_fun

Value

A numeric matrix with the predictions

Examples

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
pred <- fitted(mod)

get_loo_fun

Retrieve a loo function

Description

This function returns the correct function needed to perform one of the leave-one-out cross-validations. It’s primarily meant for internal use but can be useful when doing simulations.

Usage

get_loo_fun(x, ...)

## S4 method for signature 'tskrrHeterogeneous'
get_loo_fun(
  x,
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE
)

## S4 method for signature 'tskrrHomogeneous'
get_loo_fun(
  x,
  exclusion = c("edges", "vertices", "interaction", "both"),
  replaceby0 = FALSE
)

## S4 method for signature 'linearFilter'
get_loo_fun(x, replaceby0 = FALSE)

## S4 method for signature 'character'
get_loo_fun(
  x = c("tskrrHeterogeneous", "tskrrHomogeneous", "linearFilter"),
  ...
)
get_loo_fun

## S4 method for signature 'tskrrTune'
get_loo_fun(x, ...)

Arguments

x       a character value with the class or a tskrr or linearFilter object.
...      arguments passed to or from other methods.
exclusion a character value with possible values "interaction", "row", "column", "both" for heterogeneous models, and "edges", "vertices", "interaction" or "both" for homogeneous models. Defaults to "interaction". See details.
replaceby0 a logical value indicating whether the interaction should be simply removed (FALSE) or replaced by 0 (TRUE).

Details

This function can be used to select the correct loo function in a simulation or tuning algorithm, based on the model object you created. Depending on its class, the returned functions will have different arguments, so you should only use this if you know what you’re doing and after you checked the actual returned functions in loo_internal.

Using replaceby0 only makes sense if you only remove the interaction. In all other cases, this argument is ignored.

For the class tskrrHomogeneous, it doesn’t make sense to remove rows or columns. If you chose this option, the function will throw an error. Removing edges corresponds to the setting "edges" or "interaction". Removing vertices corresponds to the setting "vertices" or "both". These terms can be used interchangeably.

For the class linearFilter it only makes sense to exclude the interaction (i.e., a single cell). Therefore you do not have an argument exclusion for that method.

For the classes tskrrTune and tskrrImpute, not specifying exclusion or replaceby0 returns the used loo function. If you specify either of them, it will use the method for the appropriate model and return a new loo function.

Value

A function taking the arguments y, and possibly pred for calculating the leave-one-out cross-validation.

For class tskrrHeterogeneous, the returned function also has an argument Hk and Hg, representing the hat matrix for the rows and the columns respectively. For class tskrrHomogeneous, only the extra argument Hk is available. For class linearFilter, the extra argument is called alpha and takes the alpha vector of that model.

See Also

loo for carrying out a leave one out crossvalidation, and loo_internal for more information on the internal functions one retrieves with this one.
has_imputed_values  Getters for tskrrImpute objects

Description

The functions described here are convenience functions to get information out of a tskrrImpute object.

Usage

has_imputed_values(x)
which_imputed(x)
is_imputed(x)

Arguments

x a tskrrImpute object or an object inheriting from tskrrImpute.

Value

For has_imputed_values: a logical value indicating whether the model has imputed values. If x is not some form of a tskrr model, the function will return an error.
For which_imputed: a integer vector with the positions for which the values are imputed.
for is_imputed: a matrix of the same dimensions as the label matrix. It contains the value FALSE at positions that were not imputed, and TRUE at positions that were.

Examples

data(drugtarget)
mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
naid <- sample(length(drugTargetInteraction), 30)
drugTargetInteraction[naid] <- NA
impmod <- impute_tskrr(drugTargetInteraction, targetSim, drugSim)
has_imputed_values(mod)
has_imputed_values(impmod)

# For illustration: extract imputed values
id <- is_imputed(impmod)
fitted(impmod)[id]
hat

**Return the hat matrix of a tskrr model**

**Description**

This function returns the hat matrix or hat matrices of a tskrr model. xnet creates an S4 generic for hat and links the default method to the `hat` function of stats.

**Usage**

```r
hat(x, ...)
```

```r
## S4 method for signature 'tskrrHeterogeneous'
hat(x, which = c("row", "column"))
```

```r
## S4 method for signature 'tskrrHomogeneous'
hat(x, ...)
```

**Arguments**

- `x`: a tskrr model
- `...`: arguments passed to other methods.
- `which`: a character value with possible values "row" or "column" to indicate which should be returned. For homogeneous models, this parameter is ignored.

**Value**

the requested hat matrix of the model.

---

**impute_tskrr**

**Impute missing values in a label matrix**

**Description**

This function implements an optimization algorithm that allows imputing missing values in the label matrix while fitting a tskrr model.

**Usage**

```r
impute_tskrr(
  y,
  k,
  g = NULL,
  lambda = 0.01,
  testdim = TRUE,
```
testlabels = TRUE,
symmetry = c("auto", "symmetric", "skewed"),
keep = FALSE,
niter = 10000,
tol = sqrt(.Machine$double.eps),
start = mean(y, na.rm = TRUE),
verbose = FALSE
)

Arguments

y a label matrix
k a kernel matrix for the rows
g an optional kernel matrix for the columns
lambda a numeric vector with one or two values for the hyperparameter lambda. If two values are given, the first one is used for the k matrix and the second for the g matrix.
testdim a logical value indicating whether symmetry and the dimensions of the kernel(s) should be tested. Defaults to TRUE, but for large matrices putting this to FALSE will speed up the function.
testlabels a logical value indicating wether the row- and column names of the matrices have to be checked for consistency. Defaults to TRUE, but for large matrices putting this to FALSE will speed up the function.
symmetry a character value with the possibilities "auto", "symmetric" or "skewed". In case of a homogeneous fit, you can either specify whether the label matrix is symmetric or skewed, or you can let the function decide (option "auto").
keep a logical value indicating whether the kernel hat matrices should be stored in the model object. Doing so makes the model object quite larger, but can speed up predictions in some cases. Defaults to FALSE.
niter an integer giving the maximum number of iterations
tol a numeric value indicating the tolerance for convergence of the algorithm. It is the maximum sum of squared differences between to iteration steps.
start a numeric value indicating the value with which NA's are replaced in the first step of the algorithm. Defaults to 0.
verbose either a logical value, 1 or 2. 1 means "show the number of iterations and the final deviation", 2 means "show the deviation every 10 iterations". A value TRUE is read as 1.

Value

A tskrr model of the class tskrrImputeHeterogeneous or tskrrImputeHomogeneous depending on whether or not g has a value.
impute_tskrr.fit

Examples

data(drugtarget)

naid <- sample(length(drugTargetInteraction), 30)
drugTargetInteraction[naid] <- NA

impute_tskrr(drugTargetInteraction, targetSim, drugSim)

impute_tskrr.fit  Impute values based on a two-step kernel ridge regression

Description

This function provides an interface for the imputation of values based on a tskrr model and is the internal function used by impute_tskrr.

Usage

impute_tskrr.fit(y, Hk, Hg, naid = NULL, niter, tol, start, verbose)

Arguments

y  a label matrix
Hk  a hat matrix for the rows (see also eigen2hat on how to calculate them from an eigen decomposition)
Hg  a hat matrix for the columns. For homogeneous networks, this should be Hk again.
naid  an optional index with the values that have to be imputed, i.e. at which positions you find a NA value. It can be a vector with integers or a matrix with TRUE/FALSE values.
niter  an integer giving the maximum number of iterations
tol  a numeric value indicating the tolerance for convergence of the algorithm. It is the maximum sum of squared differences between to iteration steps.
start  a numeric value indicating the value with which NA's are replaced in the first step of the algorithm. Defaults to 0.
verbose  either a logical value, 1 or 2. 1 means "show the number of iterations and the final deviation", 2 means "show the deviation every 10 iterations". A value TRUE is read as 1.

Details

This function is mostly available for internal use. In most cases, it makes much more sense to use impute_tskrr, as that function returns an object one can work with. The function impute_tskrr.fit could be useful when doing simulations or creating fitting algorithms.
Value

a list with two elements:

• a matrix \( y \) with the imputed values filled in.
• a numeric value \( niter \) with the amount of iterations

See Also

• \texttt{impute_tskrr} for the user-level function, and
• \texttt{eigen2hat} for conversion of a eigen decomposition to a hat matrix.

Examples

data(drugtarget)

\begin{verbatim}
K <- eigen(targetSim)
G <- eigen(drugSim)

Hk <- eigen2hat(K$vectors, K$values, lambda = 0.01)
Hg <- eigen2hat(G$vectors, G$values, lambda = 0.05)

drugTargetInteraction[c(3,17,123)] <- NA
res <- impute_tskrr.fit(drugTargetInteraction, Hk, Hg,
                         niter = 1000, tol = 10e-10,
                         start = 0, verbose = FALSE)
\end{verbatim}

is_symmetric

\texttt{Test symmetry of a matrix}

Description

The function \texttt{isSymmetric} tests for symmetry of a matrix but also takes row and column names into account. This function is a toned-down (and slightly faster) version that ignores row and column names. Currently, the function only works for real matrices, not complex ones.

Usage

\texttt{is_symmetric(x, tol = 100 * .Machine$double.eps)}

Arguments

\begin{itemize}
  \item \texttt{x} a matrix to be tested.
  \item \texttt{tol} the tolerance for comparing the numbers.
\end{itemize}
is_tuned

Value

a logical value indicating whether or not the matrix is symmetric

Examples

```r
x <- matrix(1:16, ncol = 4)
is_symmetric(x)

x <- x %*% t(x)
is_symmetric(x)
```

---

is_tuned  

Getters for tskrrTune objects

Description

The functions described here are convenience functions to get information out of a tskrrTune object.

Usage

```r
is_tuned(x)
get_grid(x)
get_loss_values(x)
has_onedim(x)
```

Arguments

x  

a tskrrTune object or an object inheriting from tskrrTune.

Value

For is_tuned: a logical value indicating whether the model is tuned.

For get_grid a list with the elements k and possibly g, each containing the different lambdas tried in the tuning for the row and column kernel matrices respectively.

For get_loss_values a matrix with the calculated loss values. Note that each row represents the result for one lambda value related to the row kernel matrix K. For heterogeneous models, every column represents the result for one lambda related to the column kernel matrix G.

For is_onedim a single logical value telling whether the grid search in the object was onedimensional.
Examples

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
tuned <- tune(mod, ngrid = 10)

is_tuned(mod)
is_tuned(tuned)

# Basic visualization of the grid.

gridvals <- get_grid(tuned)
z <- get_loss_values(tuned)

## Not run:
image(gridvals$k, gridvals$g, log(z), log = 'xy',
      xlab = "lambda k", ylab = "lambda g")

## End(Not run)

labels.tskrr 

Extract labels from a tskrr object

Description

These functions allow you to extract the labels from a tskrr object. The function labels and the function dimnames are aliases and do the exact same thing. The functions rownames and colnames work like you would expect. Note that contrary to the latter two, labels will never return NULL. If no labels are found, it will construct labels using the prefixes defined in the argument prefix.

Usage

## S3 method for class 'tskrr'
labels(
  object,
  prefix = if (is_homogeneous(object)) "row" else c("row", "col"), ...
)

## S4 method for signature 'tskrr'
labels(
  object,
  prefix = if (is_homogeneous(object)) "row" else c("row", "col"), ...
)
linearFilter-class

## S4 method for signature 'tskrr'

### dimnames(x)

## S4 method for signature 'tskrr'

### rownames(x, do.NULL = TRUE, prefix = "row")

## S4 method for signature 'tskrr'

### colnames(x, do.NULL = TRUE, prefix = "col")

### Arguments

- **object**
  - a `tskrr` object

- **prefix**
  - a prefix used for construction of the labels in case none are available. For `label`, a character vector of length 1 for homogeneous networks or of length 2 for heterogeneous networks. In case two values are given, the first is used for the rows and the second for the columns. Otherwise the only value is used for both. In the case of `rownames` and `colnames`, a single value. See also `row+colnames`

- **...**
  - arguments passed to/from other methods.

- **x**
  - a `tskrr` object

- **do.NULL**
  - logical. If FALSE and labels are NULL, labels are created. If TRUE, the function returns NULL in the absence of labels.

### Value

- for `labels` and `dimnames`: a list with two elements k and g

### Warning

If the original data didn’t contain row- or column names for the label matrix, `rownames` and `colnames` will return NULL. Other functions will extract the automatically generated labels, so don’t count on `rownames` and `colnames` if you want to predict output from other functions!

---

linearFilter-class   Class linearFilter

### Description

The class represents the outcome of a linear filter, and is normally generated by the function `linear_filter`

### Slots

- **y** the original label matrix with responses.
- **alpha** a numeric vector with the 4 alpha values of the model.
- **pred** a matrix with the predictions
- **mean** a numeric vector containing the global mean of y
linear_filter

Fit a linear filter over a label matrix

Description

This function fits a linear filter over a label matrix. It calculates the row, column and total means, and uses those to construct the linear filter.

Usage

linear_filter(y, alpha = 0.25, na.rm = FALSE)

Arguments

y a label matrix
alpha a vector with 4 alpha values, or a single alpha value which then is used for all 4 alphas.
na.rm a logical value indicating whether missing values should be removed before calculating the row-, column- and total means.

Details

If there are missing values and they are removed before calculating the means, a warning is issued. If na.rm = FALSE and there are missing values present, the outcome is, by definition, a matrix filled with NA values.

Value

an object of class linearFilter

Examples

data(drugtarget)
linear_filter(drugTargetInteraction, alpha = 0.25)
linear_filter(drugTargetInteraction, alpha = c(0.1,0.1,0.4,0.4))

See Also

linear_filter for creating a linear filter model, and getter functions for linearFilter.
Description


Usage

```r
loo(x, ...) 
```

## S4 method for signature 'tskrrHeterogeneous'

```r
loo(
  x,
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE
)
```

## S4 method for signature 'tskrrHomogeneous'

```r
loo(
  x,
  exclusion = c("edges", "vertices", "interaction", "both"),
  replaceby0 = FALSE
)
```

## S4 method for signature 'linearFilter'

```r
loo(x, replaceby0 = FALSE)
```

Arguments

- **x**: an object of class `tskrr` or `linearFilter`.
- **...**: arguments passed to methods. See Details.
- **exclusion**: a character value with possible values "interaction", "row", "column", "both" for heterogeneous models, and "edges", "vertices", "interaction" or "both" for homogeneous models. Defaults to "interaction". See details.
- **replaceby0**: a logical value indicating whether the interaction should be simply removed (FALSE) or replaced by 0 (TRUE).

Details

The parameter `exclusion` defines what is left out. The value "interaction" means that a single interaction is removed. In the case of a homogeneous model, this can be interpreted as the removal of the interaction between two edges. The values "row" and "column" mean that all interactions for a row edge resp. a column edge are removed. The value "both" removes all interactions for a row and a column edge.
In the case of a homogeneous model, "row" and "column" don't make sense and will be replaced by "both" with a warning. This can be interpreted as removing vertices, i.e. all interactions between one edge and all other edges. Alternatively one can use "edges" to remove edges and "vertices" to remove vertices. In the case of a homogeneous model, the setting "edges" translates to "interaction", and "vertices" translates to "both". For more information, see Stock et al. (2018).

Replacing by 0 only makes sense when exclusion = "interaction" and the label matrix contains only 0 and 1 values. The function checks whether the conditions are fulfilled and if not, returns an error.

**Value**

a numeric matrix with the leave-one-out predictions for the model.

**Examples**

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim, 
lambda = c(0.01,0.01))

delta <- loo(mod, exclusion = 'both') - response(mod) 
delta0 <- loo(mod, replaceby0 = TRUE) - response(mod)

---

**Description**

These functions implement different cross-validation scenarios for two-step kernel ridge regression. It uses the shortcuts for leave-one-out cross-validation.

**Usage**

loo.i(Y, Hk, Hg, pred)

loo.i0(Y, Hk, Hg, pred)

loo.r(Y, Hk, Hg, ...)

loo.c(Y, Hk, Hg, ...)

loo.b(Y, Hk, Hg, ...)

loo.e.sym(Y, Hk, pred)

loo.e.skew(Y, Hk, pred)
loss

\begin{align*}
\text{loo.e0.sym}(Y, H_k, \text{pred}) \\
\text{loo.e0.skew}(Y, H_k, \text{pred}) \\
\text{loo.v}(Y, H_k, \ldots) \\
\text{loo.i.lf}(Y, \alpha, \text{pred}) \\
\text{loo.i0.lf}(Y, \alpha, \text{pred})
\end{align*}

Arguments

- \text{Y} \quad \text{the matrix with responses}
- \text{H}_k \quad \text{the hat matrix for the first kernel (rows of } Y\text{)}
- \text{H}_g \quad \text{the hat matrix for the second kernel (columns of } Y\text{)}
- \text{pred} \quad \text{the predictions}
- \ldots \quad \text{added to allow for specifying } \text{pred} \text{ even when not needed.}
- \alpha \quad \text{a vector of length } 4 \text{ with the alpha values from a linearFilter model}

Details

These functions are primarily for internal use and hence not exported. Be careful when using them, as they do not perform any sanity check on the input. It is up to the user to make sure the input makes sense.

Value

a matrix with the leave-one-out predictions

See Also

\texttt{loo} for the user-level function.

\begin{center}
\begin{tabular}{l}
\texttt{loss} \\
\textit{Calculate or extract the loss of a tskrr model}
\end{tabular}
\end{center}

Description

This function allows calculating the loss of a tskrr model using either one of the functions defined in \texttt{loss_functions} or a custom user function. If the model inherits from class \texttt{tskrrTune} and no additional arguments are given, the loss is returned for the settings used when tuning. The function can also be used to extract the original loss from a \texttt{permtest} object.
Usage

loss(x, ...)

## S4 method for signature 'tskrr'
loss(
  x,
  fun = loss_mse,
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE,
  predictions = FALSE,
  ...
)

## S4 method for signature 'tskrrTune'
loss(
  x,
  fun = loss_mse,
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE,
  predictions = FALSE,
  ...
)

## S4 method for signature 'permtest'
loss(x, ...)

Arguments

x a model that inherits from class tskrr
...
extra arguments passed to the loss function in fun.
fun a function to be used for calculating the loss. This can also be a character value
giving the name of one of the loss functions provided in the package
exclusion a character value with possible values "interaction", "row", "column" or "both".
See also loo for more information.
replaceby0 a logical value indicating whether the interaction should be simply removed
(FALSE) or replaced by 0 (TRUE).
predictions a logical value to indicate whether the predictions should be used instead of
leave one out crossvalidation. If set to TRUE, the other arguments are ignored.

Value

a numeric value with the calculated loss

See Also

- loss_functions for possible loss functions
- tune for tuning a model based on loss functions
Examples

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)

loss(mod, fun = loss_auc)

tuned <- tune(mod, fun = loss_auc)

loss(tuned)

loss(tuned, fun = loss_mse)

Description

These functions can be used as loss functions in tune. Currently, two functions are provided: a function calculating the classic mean squared error (loss_mse) and a function calculating 1 - AUC (loss_auc).

Usage

loss_mse(Y, LOO, na.rm = FALSE)

loss_auc(Y, LOO)

Arguments

Y the label matrix with observed responses

LOO the leave-one-out crossvalidation (or predictions if you must). This one can be calculated by the function loo.

na.rm a logical value

Details

The AUC is calculated by sorting the Y matrix based on the order of the values in the LOO matrix. The false and true positive rates are calculated solely based on that ordering, which allows for values in LOO outside the range [0,1]. It’s a naive implementation which is good enough for tuning, but shouldn’t be used as a correct value for 1 - auc in case the values in LOO are outside the range [0,1].

Note

The function loss_auc should only be used for a Y matrix that contains solely the values 0 and 1.
See Also
tune for application of the loss function

Examples

```r
x <- c(1,0,0,1,0,0,1,0,1)
y <- c(0.8,-0.1,0.2,0.2,0.4,0.01,1.12,0.9,0.9)
loss_mse(x,y)
loss_auc(x,y)
```

---

match_labels Reorder the label matrix

Description

Reorders the label matrix based on the labels of the kernel matrices. In case there are no labels, the original label matrix is returned, but with the labels in rows and cols as rownames and column names respectively.

Usage

```r
match_labels(y, rows, cols = NULL)
```

Arguments

- `y` a matrix representing the label matrix.
- `rows` a character vector with the labels for the rows or a matrix with rownames that will be used as labels.
- `cols` a character vector with the labels for the cols or a matrix with colnames that will be used as labels. If NULL, rows will be used for both row and column labels.

Value

a matrix with the rows and columns reordered.

Examples

```r
mat <- matrix(1:6, ncol = 2,
              dimnames = list(c("b", "a", "d"),
                              c("ca", "cb")))
match_labels(mat, c("a","b", "d"), c("ca","cb"))

#Using matrices
data(drugtarget)
out <- match_labels(drugTargetInteraction, targetSim, drugSim)
```
permtest  Calculate the relative importance of the edges

Description

This function does a permutation-based evaluation of the impact of different edges on the final result. It does so by permuting the kernel matrices, refitting the model and calculating a loss function.

Usage

permtest(x, ...)

## S3 method for class 'permtest'
print(x, digits = max(3L, getOption("digits") - 3), ...)

## S4 method for signature 'tskrrHeterogeneous'
permtest(
  x,
  n = 100,
  permutation = c("both", "row", "column"),
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE,
  fun = loss_mse,
  exact = FALSE
)

## S4 method for signature 'tskrrHomogeneous'
permtest(
  x,
  n = 100,
  permutation = c("both"),
  exclusion = c("interaction", "both"),
  replaceby0 = FALSE,
  fun = loss_mse,
  exact = FALSE
)

## S4 method for signature 'tskrrTune'
permtest(x, permutation = c("both", "row", "column"), n = 100)

Arguments

x        either a tskrr-class or a tskrrTune-class object
...
        arguments passed to other methods
digits    the number of digits shown in the output
n        the number of permutations for every kernel matrix
permtest-class

description of the permtest-class

Details

The test involved uses a normal approximation. It assumes that under the null hypothesis, the loss values are approximately normally distributed. The cumulative probability of a loss as small or smaller than the one found in the original model, is calculated based on a normal distribution from which the mean and sd are calculated from the permutations.

Value

An object of the class permtest.

Warning

It should be noted that this normal approximation is an ad-hoc approach. There’s no guarantee that the actual distribution of the loss under the null hypothesis is normal. Depending on the loss function, a significant deviation from the theoretic distribution can exist. Hence this functions should only be used as a rough guidance in model evaluation.

Examples

# Heterogeneous network
data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
permtest(mod, fun = loss_auc)
permutations 29

Slots

orig.loss a numeric value with the original loss of the model.
perm.losses a numeric vector with the losses of the different permutations.
n the number of permutations
loss.function the function used to calculate the losses.
exclusion a character value indicating the exclusion setting used for the test
replaceby0 a logical value that indicates whether the exclusion was done by replacing with zero.
See also loo.
permutation a character value that indicates in which kernel matrices were permuted.
pval a p value indicating how likely it is to find a smaller loss than the one of the model based on a normal approximation.
extact a logical value indicating whether the P value was calculated exactly or approximated by the normal distribution.

See Also

- the function permtest for the actual test.
- the function loo for the leave one out procedures
- the function t.test for the actual test

permutations (Getters for permtest objects)

Description

The functions described here are convenience functions to get information out of a permtest object.

Usage

permutations(x)

## S4 method for signature 'permtest'
x[i]

Arguments

x a permtest object
i either a numeric vector, a logical vector or a character vector with the elements that need extraction.

Value

the requested values
See Also

`loss` to extract the original loss value.

Examples

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
ptest <- permtest(mod, fun = loss_auc)

loss(ptest)
ptest[c(2,3)]
permutations(ptest)

---

plot.tskrr

`plot a heatmap of the predictions from a tskrr model`

Description

This function plots a heatmap of the fitted values in a tskrr model. The function is loosely based on `heatmap`, but uses a different mechanism and adds a legend by default.

Usage

```r
## S3 method for class 'tskrr'
plot(
  x, dendro = c("both", "row", "col", "none"),
  which = c("fitted", "loo", "response", "residuals"),
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE,
  nbest = 0, rows, cols,
  col = rev(heat.colors(20)),
  breaks = NULL, legend = TRUE,
  main = NULL, xlab = NULL, ylab = NULL,
  labRow = NULL, labCol = NULL,
  margins = c(5, 5),
  ...
)
```
interrupt

Arguments

- **x**
  - A `tskr` model

- **dendro**
  - A character value indicating whether a dendrogram should be constructed.

- **which**
  - A character value indicating whether the fitted values, the leave-one-out values, the original response values or the residuals should be plotted.

- **exclusion**
  - If `which = "loo"`, this argument is passed to `loo` for the exclusion settings.

- **replaceby0**
  - If `which = "loo"`, this argument is passed to `loo`.

- **nbest**
  - A single integer value indicating the amount of best values that should be selected. If 0, all data is shown.

- **rows**
  - A numeric or character vector indicating which rows should be selected from the model.

- **cols**
  - A numeric or character vector indicating which columns should be selected from the model.

- **col**
  - A vector with colors to be used for plotting.

- **breaks**
  - A single value specifying the number of breaks (must be 1 more than number of colors), or a numeric vector with the breaks used for the color code. If `NULL`, the function tries to find evenly spaced breaks.

- **legend**
  - A logical value indicating whether or not the legend should be added to the plot.

- **main**
  - A character value with a title for the plot.

- **xlab**
  - A character label for the X axis.

- **ylab**
  - A character label for the Y axis.

- **labRow**
  - A character vector with labels to be used on the rows. Note that these labels are used as is (possibly reordered to match the dendrogram). They can replace the labels from the model. Set to `NA` to remove the row labels.

- **labCol**
  - The same as `labRow` but then for the columns.

- **margins**
  - A numeric vector with 2 values indicating the margins to be used for the row and column labels (cfr `par("mar")`)

Details

The function can select a part of the model for plotting. Either you specify `rows` and `cols`, or you specify `nbest`. If `nbest` is specified, `rows` and `cols` are ignored. The n highest values are looked up in the plotted values, and only the rows and columns related to these values are shown then. This allows for a quick selection of the highest predictions.

Dendrograms are created by converting the kernel matrices to a distance, using

\[ d(x,y) = K(x,x)^2 + K(y,y)^2 - 2*K(x,y) \]

with K being the kernel function. The resulting distances are clustered using `hclust` and converted to a dendrogram using `as.dendrogram`. 

... currently ignored...
Value

an invisible list with the following elements:

- val: the values plotted
- ddK: if a row dendrogram was requested, the row dendrogram
- ddG: if a column dendrogram was requested, the column dendrogram
- breaks: the breaks used for the color codes
- col: the colors used

See Also
tskrr, tune and link{impute_tskrr} to construct tskrr models.

Examples

data(drugtarget)
mod <- tskrr(drugTargetInteraction, targetSim, drugSim)

plot(mod)
plot(mod, dendro = "row", legend = FALSE)
plot(mod, col = rainbow(20), dendro = "none", which = "residuals")
plot(mod, labCol = NA, labRow = NA, margins = c(0.2,0.2))

plot_grid

Plot the grid of a tuned tskrr model

Description

With this function, you can visualize the grid search for optimal lambdas from a tskrrTune object. In the case of two-dimensional grid search, this function plots a contour plot on a grid, based on the functions image and contour. For one-dimensional grid search, the function creates a single line plot.

Usage

plot_grid(
x,
addlambda = TRUE,
lambdapars = list(col = "red"),
log = TRUE,
opts.contour = list(nlevels = 10),
...
predict.tskrr

Arguments

- **x**: an object that inherits from `tskrrTune`
- **addlambda**: a logical value indicating whether the lambda with the minimum loss should be added to the plot. In case of a one dimensional plot, this adds a colored vertical line. In the case of a two dimensional plot, this adds a colored point at the minimum.
- **lambdapars**: a list with named `par` values passed to the function `abline` or `points` for plotting the best lambda value when `addmin = TRUE`.
- **log**: a logical value indicating whether the lambdas should be plotted at a log scale (the default) or not.
- **opts.contour**: options passed to the function `contour` for 2D grid plots. Ignored for 1D grid plots.
- **...**: arguments passed to other functions. For a one dimensional plot, this will be the function `plot`

Value

- `NULL` invisibly

Examples

```r
data(drugtarget)

## One dimensional tuning
 tuned1d <- tune(drugTargetInteraction, targetSim, drugSim,
                 lim = c(1e-4, 2), ngrid = 40,
                 fun = loss_auc, onedim = TRUE)

plot_grid(tuned1d)
plot_grid(tuned1d, lambdapars = list(col = "green",
                                     lty = 1, lwd = 2),
          log = FALSE, las = 2, main = "1D tuning")

## Two dimensional tuning
 tuned2d <- tune(drugTargetInteraction, targetSim, drugSim,
                lim = c(1e-4, 10), ngrid = 20,
                fun = loss_auc)

plot_grid(tuned2d)
```

---

**predict.tskrr**

*predict method for tskrr fits*
Description

Obtains the predictions from a tskrr model for new data. To get the predictions on the training data, use the function fitted or set both k and g to NULL.

Usage

```r
## S3 method for class 'tskrr'
predict(object, k = NULL, g = NULL, testdim = TRUE, ...)

## S4 method for signature 'tskrr'
predict(object, k = NULL, g = NULL, testdim = TRUE, ...)
```

Arguments

- **object**: an object of class tskrr.
- **k**: a new K matrix or NULL. If NULL, the fitted values on the training data are returned.
- **g**: a new G matrix or NULL. If NULL, K is used for both.
- **testdim**: a logical value indicating whether the dimensions should be checked prior to the calculation. You can set this to FALSE but you might get more obscure errors if dimensions don’t match.
- **...**: arguments passed to or from other methods

Details

Predictions can be calculated between new vertices and the vertices used to train the model, between new sets of vertices, or both. Which predictions are given, depends on the kernel matrices passed to the function.

In any case, both the K and G matrix need the kernel values for every combination of the new vertices and the vertices used to train the model. This is illustrated for both homogeneous and heterogeneous networks in the examples.

To predict the links between a new set of vertices and the training vertices, you need to provide the kernel matrix for either the K or the G set of vertices. If you want to predict the mutual links between two new sets of vertices, you have to provide both the K and the G matrix. This is particularly important for homogeneous networks: if you only supply the k argument, you will get predictions for the links between the new vertices and the vertices on which the model is trained. So in order to get the mutual links between the new vertices, you need to provide the kernel matrix as the value for both the k and the g argument.

Value

- a matrix with predicted values.
Warning

This function is changed in version 0.1.9 so it’s more consistent in how it expects the K and G matrices to be ordered. Up to version 0.1.8 the new vertices should be on the rows for the K matrix and on the columns for the G matrix. This leads to confusion.

If you’re using old code, you’ll get an error pointing this out. You need to transpose the G matrix in the old code to make it work with the new version.

See Also

tskrr and tskrrTune for fitting the models.

Examples

### Predictions for homogeneous networks

data(proteinInteraction)

idnew <- sample(nrow(Kmat_y2h_sc), 20)

trainY <- proteinInteraction[-idnew,-idnew]
trainK <- Kmat_y2h_sc[-idnew,-idnew]

testK <- Kmat_y2h_sc[idnew, - idnew]

mod <- tskrr(trainY, trainK, lambda = 0.1)
# Predict interaction between test vertices
predict(mod, testK, testK)

# Predict interaction between test and train vertices
predict(mod, testK)
predict(mod, g = testK)

### Predictions for heterogeneous networks

data("drugtarget")

idnewK <- sample(nrow(targetSim), 10)
idnewG <- sample(ncol(drugSim), 10)

trainY <- drugTargetInteraction[-idnewK, -idnewG]
trainK <- targetSim[-idnewK, -idnewK]
trainG <- drugSim[-idnewG, -idnewG]

testK <- targetSim[idnewK, -idnewK]
testG <- drugSim[idnewG, -idnewG]

mod <- tskrr(trainY, trainK, trainG, lambda = 0.01)

# Predictions for new targets on drugs in model
predict(mod, testK)
# Predictions for new drugs on targets in model
predict(mod, g = testG)
# Predictions for new drugs and targets
predict(mod, testK, testG)

proteinInteraction      Protein interaction for yeast

Description
A dataset for examining the interaction between proteins of yeast. The dataset consists of the following objects:

Usage
proteinInteraction

Format
- proteinInteraction: a numeric square matrix with 150 rows/columns
- Kmat_y2h_sc: a numeric square matrix with 150 rows/columns

Details
- proteinInteraction: the label matrix based on the protein network taken from the KEGG/PATHWAY database
- Kmat_y2h_sc: a kernel matrix indicating similarity of proteins.

The proteins in the dataset are a subset of the 769 proteins used in Yamanishi et al (2004). The kernel matrix used is the combination of 4 kernels: one based on expression data, one on protein interaction data, one on localization data and one on phylogenetic profile. These kernels and their combination are also explained in Yamanishi et al (2004).

Source
https://doi.org/10.1093/bioinformatics/bth910

References
Yamanishi et al, 2004: Protein network inference from multiple genomic data: a supervised approach.
residuals

calculate residuals from a tskrr model

Description

This function returns the residuals for an object inheriting from class tskrr

Usage

residuals(object, ...)

## S3 method for class 'tskrr'
residuals(
  object,
  method = c("predictions", "loo"),
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE,
  ...
)

## S4 method for signature 'tskrr'
residuals(
  object,
  method = c("predictions", "loo"),
  exclusion = c("interaction", "row", "column", "both"),
  replaceby0 = FALSE,
  ...
)

Arguments

object          a tskrr model
...             arguments passed from/to other methods.
method          a character value indicating whether the residuals should be based on the predictions or on a leave-one-out crossvalidation.
exclusion       a character value with possible values "interaction", "row", "column", "both" for heterogeneous models, and "edges", "vertices", "interaction" or "both" for homogeneous models. Defaults to "interaction". See details.
replaceby0      a logical value indicating whether the interaction should be simply removed (FALSE) or replaced by 0 (TRUE).

Details

The parameter exclusion defines what is left out. The value "interaction" means that a single interaction is removed. In the case of a homogeneous model, this can be interpreted as the removal of the interaction between two edges. The values "row" and "column" mean that all interactions for
a row edge resp. a column edge are removed. The value "both" removes all interactions for a row
and a column edge.

In the case of a homogeneous model, "row" and "column" don't make sense and will be replaced
by "both" with a warning. This can be interpreted as removing vertices, i.e. all interactions between
one edge and all other edges. Alternatively one can use "edges" to remove edges and "vertices" to
remove vertices. In the case of a homogeneous model, the setting "edges" translates to "interaction",
and "vertices" translates to "both". For more information, see Stock et al. (2018).

Replacing by 0 only makes sense when exclusion = "interaction" and the label matrix contains
only 0 and 1 values. The function checks whether the conditions are fulfilled and if not, returns an
error.

Value

a matrix(!) with the requested residuals

Examples

data(drugtarget)
mod <- tskrr(drugTargetInteraction, targetSim, drugSim,
  lambda = c(0.01,0.01))
delta <- response(mod) - loo(mod, exclusion = "both")
resid <- residuals(mod, method = "loo", exclusion = "both")
all.equal(delta, resid)

Description

The functions described here are convenience functions to get information out of a tskrr
object.

Usage

## S4 method for signature 'tskrr'
response(x, ...)

## S4 method for signature 'tskrrHomogeneous'
lambda(x)

## S4 method for signature 'tskrrHeterogeneous'
lambda(x)

## S4 method for signature 'tskrrHomogeneous'
lambda(x)

is_tskrr(x)

is_homogeneous(x)
is_heterogeneous(x)

symmetry(x)

get_eigen(x, which = c("row", "column"))

get_kernelmatrix(x, which = c("row", "column"))

has_hat(x)

get_kernel(x, which = c("row", "column"))

Arguments

x a tskrr object or an object inheriting from tskrr.

... arguments passed to other methods.

which a character value indicating whether the eigen decomposition for the row kernel matrix or the column kernel matrix should be returned.

Value

For response: the original label matrix

For lambda: a named numeric vector with one resp both lambda values used in the model. The names are "k" and "g" respectively.

For is_tskrr a logical value indicating whether the object is a tskrr object

For is_homogeneous a logical value indicating whether the tskrr model is a homogeneous one.

For is_heterogeneous a logical value indicating whether the tskrr model is a heterogeneous one.

For symmetry a character value indicating the symmetry for a homogeneous model. If the model is not homogeneous, NA is returned.

For get_eigen the eigen decomposition of the requested kernel matrix.

For get_kernelmatrix the original kernel matrix for the rows or columns.

For has_hat a logical value indicating whether the tskrr model contains the kernel hat matrices.

Warning

The function get_kernel is deprecated. Use get_kernelmatrix instead.

Examples

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)

is_homogeneous(mod)

EigR <- get_eigen(mod)

EigC <- get_eigen(mod, which = 'column')
test_symmetry

Description

This function tells you whether a matrix is symmetric, skewed symmetric, or not symmetric. It’s used by tskrr to determine which kind of homologous network is represented by the label matrix.

Usage

test_symmetry(x, tol = .Machine$double.eps)

Arguments

x   a matrix

 tol  a single numeric value with the tolerance for comparison

Value

a character value with the possible values "symmetric", "skewed" or "none".

See Also

 tskrrHomogeneous for more information on the values for the slot symmetry

Examples

mat1 <- matrix(c(1,0,0,1),ncol = 2)
test_symmetry(mat1)
mat2 <- matrix(c(1,0,0,-1), ncol = 2)
test_symmetry(mat2)
mat3 <- matrix(1:4, ncol = 2)
test_symmetry(mat3)
tskrr is the primary function for fitting a two-step kernel ridge regression model. It can be used for both homogeneous and heterogeneous networks.

```r
  tskrr(y, k, g = NULL, lambda = 1e-04, testdim = TRUE, testlabels = TRUE, symmetry = c("auto", "symmetric", "skewed"), keep = FALSE)
```

**Arguments**
- `y`: a label matrix
- `k`: a kernel matrix for the rows
- `g`: an optional kernel matrix for the columns
- `lambda`: a numeric vector with one or two values for the hyperparameter lambda. If two values are given, the first one is used for the `k` matrix and the second for the `g` matrix.
- `testdim`: a logical value indicating whether symmetry and the dimensions of the kernel(s) should be tested. Defaults to `TRUE`, but for large matrices putting this to `FALSE` will speed up the function.
- `testlabels`: a logical value indicating whether the row- and column names of the matrices have to be checked for consistency. Defaults to `TRUE`, but for large matrices putting this to `FALSE` will speed up the function.
- `symmetry`: a character value with the possibilities "auto", "symmetric" or "skewed". In case of a homogeneous fit, you can either specify whether the label matrix is symmetric or skewed, or you can let the function decide (option "auto").
- `keep`: a logical value indicating whether the kernel hat matrices should be stored in the model object. Doing so makes the model object quite larger, but can speed up predictions in some cases. Defaults to `FALSE`.

**Value**
- a `tskrr` object
### See Also

- `response`, `fitted`, `get_eigen`, `eigen2hat`

### Examples

```r
# Heterogeneous network
data(drugtarget)
mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
Y <- response(mod)
pred <- fitted(mod)

# Homogeneous network
data(proteinInteraction)
modh <- tskrr(proteinInteraction, Kmat_y2h_sc)
Yh <- response(modh)
pred <- fitted(modh)
```

---

### Description

The class `tskrr` represents a two step kernel ridge regression fitting object, and is normally generated by the function `tskrr`. This is a superclass so it should not be instantiated directly.

### Slots

- `y` the matrix with responses
- `k` the eigen decomposition of the kernel matrix for the rows
- `lambda.k` the lambda value used for `k`
- `pred` the matrix with the predictions
- `has.hat` a logical value indicating whether the kernel hat matrices are stored in the object.
- `Hk` the kernel hat matrix for the rows.
- `labels` a list with two character vectors, `k` and `g`, containing the labels for the rows resp. columns.

See `tskrrHomogeneous` and `tskrrHeterogeneous` for more details.

### See Also

- the classes `tskrrHomogeneous` and `tskrrHeterogeneous` for the actual classes.
tskrr.fit

---

*tskrr.fit*    
*Carry out a two-step kernel ridge regression*

---

**Description**

This function provides an interface for two-step kernel ridge regression. To use this function, you need at least one kernel matrix and one label matrix. It’s the internal engine used by the function *tskrr*.

**Usage**

```r
 tskrr.fit(y, k, g = NULL, lambda.k = NULL, lambda.g = NULL, ...)
```

**Arguments**

- `y`  
  a matrix representing the links between the nodes of both networks.

- `k`  
  an object of class `eigen` containing the eigen decomposition of the first kernel matrix.

- `g`  
  an optional object of class `eigen` containing the eigen decomposition of the second kernel matrix. If `NULL`, the network is considered to be homogeneous.

- `lambda.k`  
  a numeric value for the lambda parameter tied to the first kernel.

- `lambda.g`  
  a numeric value for the lambda parameter tied to the second kernel. If `NULL`, the model is fit using the same value for `lambda.k` and `lambda.g`

- `...`  
  arguments passed to other functions. Currently ignored.

**Details**

This function is mostly available for internal use. In most cases, it makes much more sense to use *tskrr*, as that function returns an object one can work with. The function `tskrr.fit` could be useful when doing simulations or fitting algorithms, as the information returned from this function is enough to use the functions returned by *get_loo_fun*.

**Value**

a list with three elements:

- `k` : the hat matrix for the rows
- `g` : the hat matrix for the columns (or `NULL`) for homogeneous networks.
- `pred` : the predictions
Examples

```r
data(drugTarget)
K <- eigen(targetSim)
G <- eigen(drugSim)
res <- tskrr.fit(drugTargetInteraction,K,G,
    lambda.k = 0.01, lambda.g = 0.05)
```

---

### tskrrHeterogeneous-class

**Class tskrrHeterogeneous**

---

#### Description

The class tskrrHeterogeneous is a subclass of the superclass tskrr specifically for heterogeneous networks.

#### Slots

- **y** the matrix with responses
- **k** the eigen decomposition of the kernel matrix for the rows
- **lambda.k** the lambda value used for k
- **pred** the matrix with the predictions
- **g** the eigen decomposition of the kernel matrix for the columns
- **lambda.g** the lambda value used for g
- **has.hat** a logical value indicating whether the kernel hat matrices are stored in the object.
- **Hk** the kernel hat matrix for the rows.
- **Hg** the kernel hat matrix for the columns.
- **labels** a list with elements k and g (see tskrr-class). If any element is NA, the labels used are integers indicating the row resp column number.
Class tskrrHomogeneous

Description

The class tskrrHomogeneous is a subclass of the superclass tskrr specifically for homogeneous networks.

Slots

- `y` the matrix with responses
- `k` the eigen decomposition of the kernel matrix for the rows
- `lambda.k` the lambda value used for k
- `pred` the matrix with the predictions
- `symmetry` a character value that can have the possible values "symmetric", "skewed" or "not". It indicates whether the `y` matrix is symmetric, skewed-symmetric or not symmetric.
- `has.hat` a logical value indicating whether the kernel hat matrices are stored in the object.
- `Hk` the kernel hat matrix for the rows.
- `labels` a list with elements `k` and `g` (see tskrr-class). For homogeneous networks, `g` is always `NA`. If `k` is `NA`, the labels used are integers indicating the row resp column number.

Class tskrrImpute

Description

The class tskrrImpute is a virtual class that represents a tskrr model with imputed values in the label matrix Y. Apart from the model, it contains the following extra information on the imputed values.

Slots

- `imputeid` a vector with integer values indicating which of the values in `y` are imputed
- `niter` an integer value giving the number of iterations used
- `tol` a numeric value with the tolerance used
Class tskrrImputeHeterogeneous

Description

The class tskrrImputeHeterogeneous is a subclass of the class tskrrHeterogeneous and tskrrImpute specifically for heterogeneous networks with imputed values. It is the result of the function impute_tskrr.

Slots

y the matrix with responses
k the eigen decomposition of the kernel matrix for the rows
lambda.k the lambda value used for k
pred the matrix with the predictions
g the eigen decomposition of the kernel matrix for the columns
lambda.g the lambda value used for g
has.hat a logical value indicating whether the kernel hat matrices are stored in the object.
Hk the kernel hat matrix for the rows.
Hg the kernel hat matrix for the columns.
labels a list with elements k and g (see tskrr-class). If any element is NA, the labels used are integers indicating the row resp column number.
imputeid a vector with integer values indicating which of the values in y are imputed
niter an integer value giving the number of iterations used
tol a numeric value with the tolerance used

Class tskrrImputeHomogeneous

Description

The class tskrrImputeHomogeneous is a subclass of the class tskrrHomogeneous and tskrrImpute specifically for homogeneous networks with imputed values. It is the result of the function impute_tskrr on a homogeneous network model.
Slots

- `y` the matrix with responses
- `k` the eigen decomposition of the kernel matrix for the rows
- `lambda.k` the lambda value used for k
- `pred` the matrix with the predictions
- `symmetry` a character value that can have the possible values "symmetric", "skewed" or "not". It indicates whether the y matrix is symmetric, skewed-symmetric or not symmetric.
- `has.hat` a logical value indicating whether the kernel hat matrices are stored in the object.
- `Hk` the kernel hat matrix for the rows.
- `labels` a list with elements k and g (see tskrr-class). For homogeneous networks, g is always NA. If k is NA, the labels used are integers indicating the row resp column number.
- `imputeid` a vector with integer values indicating which of the values in y are imputed
- `niter` an integer value giving the number of iterations used
- `tol` a numeric value with the tolerance used

### tskrrTune-class

**Class** tskrrTune

**Description**

The class tskrrTune represents a tuned tskrr model, and is the output of the function `tune`. Apart from the model, it contains extra information on the tuning procedure. This is a virtual class only.

**Slots**

- `lambda_grid` a list object with the elements k and possibly g indicating the tested lambda values for the row kernel K and - if applicable - the column kernel G. Both elements have to be numeric.
- `best_loss` a numeric value with the loss associated with the best lambdas
- `loss_values` a matrix with the loss results from the searched grid. The rows form the X dimension (related to the first lambda), the columns form the Y dimension (related to the second lambda if applicable)
- `loss_function` the used loss function
- `exclusion` a character value describing the exclusion used
- `replaceby0` a logical value indicating whether or not the cross validation replaced the excluded values by zero
- `onedim` a logical value indicating whether the grid search was done in one dimension. For homogeneous networks, this is true by default.

**See Also**

- the function `tune` for the tuning itself
- the class `tskrrTuneHomogeneous` and `tskrrTuneHeterogeneous` for the actual classes.
tskrrTuneHeterogeneous-class

Class tskrrTuneHeterogeneous

Description
The class tskrrTuneHeterogeneous represents a tuned Heterogeneous tskrr model. It inherits from the classes tskrrHeterogeneous and tskrrTune.

tskrrTuneHomogeneous-class

Class tskrrTuneHomogeneous

Description
The class tskrrTuneHomogeneous represents a tuned homogeneous tskrr model. It inherits from the classes tskrrHomogeneous and tskrrTune.

tune

tune the lambda parameters for a tskrr

Description
This function lets you tune the lambda parameter(s) of a two-step kernel ridge regression model for optimal performance. You can either tune a previously fitted tskrr model, or pass the label matrix and kernel matrices to fit and tune a model in one go.

Usage

```r
## S4 method for signature 'tskrrHomogeneous'
tune(
x,
lim = c(1e-04, 1),
ngrid = 10,
lambda = NULL,
fun = loss_mse,
exclusion = "edges",
replaceby0 = FALSE,
onedim = TRUE,
...,
)
```

## S4 method for signature 'tskrrHeterogeneous'

```r
tune(\n x,\n lim = c(1e-04, 1),\n ngrid = 10,\n lambda = NULL,\n fun = loss_mse,\n exclusion = "edges",\n replaceby0 = FALSE,\n onedim = TRUE,\n ...\n)
```
tune

x,
lam = c(1e-04, 1),
ngrid = 10,
lam = NULL,
fun = loss_mse,
exclusion = "interaction",
replaceby0 = FALSE,
onedim = FALSE,
...

## S4 method for signature 'matrix'
tune()

x,
k,
g = NULL,
lam = c(1e-04, 1),
ngrid = 10,
lam = NULL,
fun = loss_mse,
exclusion = "interaction",
replaceby0 = FALSE,
testdim = TRUE,
testlabels = TRUE,
symmetry = c("auto", "symmetric", "skewed"),
keep = FALSE,
onedim = is.null(g),
...

Arguments

x a tskrr object representing a two step kernel ridge regression model.

lim a vector with 2 values that give the boundaries for the domain in which lambda is searched, or possibly a list with 2 elements. See details.

ngrid a single numeric value giving the number of points in a single dimension of the grid, or possibly a list with 2 elements. See details.

lambda a vector with the lambdas that need checking for homogeneous networks, or possibly a list with two elements for heterogeneous networks. See Details. Defaults to NULL, which means that the function constructs the search grid from the other arguments.

fun a loss function that takes the label matrix Y and the result of the crossvalidation LOO as input. The function name can be passed as a character string as well.

exclusion a character value with possible values "interaction", "row", "column", "both" for heterogeneous models, and "edges", "vertices", "interaction" or "both" for homogeneous models. Defaults to "interaction". See details.
replaceby0  a logical value indicating whether the interaction should be simply removed (FALSE) or replaced by 0 (TRUE).

onedim  a logical value indicating whether the search should be done in a single dimension. See details.

...  arguments to be passed to the loss function

k  a kernel matrix for the rows

g  an optional kernel matrix for the columns

testdim  a logical value indicating whether symmetry and the dimensions of the kernel(s) should be tested. Defaults to TRUE, but for large matrices putting this to FALSE will speed up the function.

testlabels  a logical value indicating whether the row- and column names of the matrices have to be checked for consistency. Defaults to TRUE, but for large matrices putting this to FALSE will speed up the function.

symmetry  a character value with the possibilities "auto", "symmetric" or "skewed". In case of a homogeneous fit, you can either specify whether the label matrix is symmetric or skewed, or you can let the function decide (option "auto").

keep  a logical value indicating whether the kernel hat matrices should be stored in the model object. Doing so makes the model object quite larger, but can speed up predictions in some cases. Defaults to FALSE.

Details

This function currently only performs a simple grid search for all (combinations of) lambda values. If no specific lambda values are provided, then the function uses create_grid to create an evenly spaced (on a logarithmic scale) grid.

In the case of a heterogeneous network, you can specify different values for the two parameters that need tuning. To do so, you need to provide a list with the settings for every parameter to the arguments lim, ngrid and/or lambda. If you try this for a homogeneous network, the function will return an error.

Alternatively, you can speed up the grid search by searching in a single dimension. When onedim = TRUE, the search for a heterogeneous network will only consider cases where both lambda values are equal.

The arguments exclusion and replaceby0 are used by the function get_loo_fun to find the correct leave-one-out function.

By default, the function uses standard mean squared error based on the cross-validation results as a measure for optimization. However, you can provide a custom function if needed, as long as it takes two matrices as input: Y being the observed interactions and LOO being the result of the chosen cross-validation.

Value

a model of class tskrrTune
See Also

- `loo`, `loo_internal` and `get_loo_fun` for more information on how leave one out validation works.
- `tskrr` for fitting a two-step kernel ridge regression.
- `loss_functions` for different loss functions.

Examples

data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)
tuned <- tune(mod, lim = c(0.1,1), ngrid = list(5,10),
            fun = loss_auc)

## Not run:
# This is just some visualization of the matrix
# It can be run safely.
gridvals <- get_grid(tuned)
z <- get_loss_values(tuned)  # loss values
image(gridvals$k,gridvals$g,z, log = 'xy',
      xlab = "lambda k", ylab = "lambda g",
      col = rev(heat.colors(20)))

## End(Not run)
valid_dimensions

Arguments

- **object**: a `tskrr` object
- **lambda**: a numeric vector with one or two values for the hyperparameter lambda. If two values are given, the first one is used for the k matrix and the second for the g matrix.

Value

- an updated `tskrr` object fitted with the new lambdas.

Examples

```r
data(drugtarget)

mod <- tskrr(drugTargetInteraction, targetSim, drugSim)

# Update with the same lambda
mod2 <- update(mod, lambda = 1e-3)

# Use different lambda for rows and columns
mod3 <- update(mod, lambda = c(0.01, 0.001))

# A model with the hat matrices stored
lambda <- c(0.001, 0.01)
modkeep <- tskrr(drugTargetInteraction, targetSim, drugSim, keep = TRUE)
Hk_1 <- hat(modkeep, which = "row")
modkeep2 <- update(modkeep, lambda = lambda)
Hk_2 <- hat(modkeep2, which = "row")

# Calculate new hat matrix by hand:
decomp <- get_eigen(modkeep, which = "row")
Hk_byhand <- eigen2hat(decomp$vectors, decomp$values, lambda = lambda[1])
identical(Hk_2, Hk_byhand)
```

valid_dimensions

Functions to check matrices

Description

These functions allow you to check whether the dimensions of the label matrix and the kernel matrix (matrices) are compatible. `valid_dimensions` checks whether both k and g are square matrices, whether y has as many rows as k and whether y has as many columns as g. `is_square` checks whether both dimensions are the same.
valid_labels

Usage

valid_dimensions(y, k, g = NULL)

is_square(x)

Arguments

y a label matrix
k a kernel matrix
g an optional second kernel matrix or NULL otherwise.
x any matrix

Value

a logical value indicating whether the dimensions of the matrices are compatible for a two step kernel ridge regression.

Note

The function is_square is not exported

valid_labels Test the correctness of the labels.

Description

This function checks whether the labels between the Y, K, and G matrices make sense. This means that all the labels found as rownames for y can be found as rownames and column names of k, and all the colnames for y can be found as rownames and colnames of g (if provided).

Usage

valid_labels(y, k, g = NULL)

Arguments

y the label matrix
k the kernel matrix for the rows
g the kernel matrix for the columns (optional). If not available, it takes the value NULL
Details

Compatible labels mean that it is unequivocally clear which rows and columns can be linked throughout the model. In case none of the matrices have row- or colnames, the labels are considered compatible. In all other cases, all matrices should have both row and column names. They should fulfill the following conditions:

- the row- and column names of a kernel matrix must contain the same values in the same order. Otherwise, the matrix can’t be symmetric.
- the rownames of y should correspond to the rownames of k
- the colnames of y should correspond to the colnames of g if it is supplied, or the colnames of k in case g is NULL

Value

TRUE if all labels are compatible, an error otherwise.

Note

This is a non-exported convenience function.

weights, tskrrHeterogeneous-method

Extract weights from a tskrr model

Description

This function calculates the weight matrix for calculating the predictions of a tskrr model.

Usage

```r
## S4 method for signature 'tskrrHeterogeneous'
weights(object)
```

```r
## S4 method for signature 'tskrrHomogeneous'
weights(object)
```

Arguments

- `object` a `tskrr` object for which the weights have to be calculated.

Details

The weight matrix is calculated from the map matrices through the function `eigen2map`.

Value

a matrix with the weights for the tskrr model.
Note

The package xnet adds a S4 generic function for weights.
<table>
<thead>
<tr>
<th>Topic</th>
<th>datasets</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>drugTargetInteraction, 7</td>
</tr>
<tr>
<td></td>
<td>proteinInteraction, 36</td>
</tr>
<tr>
<td></td>
<td>[,permtest-method (permutations), 29</td>
</tr>
<tr>
<td></td>
<td>abline, 33</td>
</tr>
<tr>
<td></td>
<td>alpha, 4</td>
</tr>
<tr>
<td></td>
<td>alpha,linearFilter-method (alpha), 4</td>
</tr>
<tr>
<td></td>
<td>as.dendrogram, 31</td>
</tr>
<tr>
<td></td>
<td>as.tskrr (as_tuned), 5</td>
</tr>
<tr>
<td></td>
<td>as.tskrr,tskrr-method (as_tuned), 5</td>
</tr>
<tr>
<td></td>
<td>as.tskrr,tskrrImpute-method (as_tuned), 5</td>
</tr>
<tr>
<td></td>
<td>as.tskrr,tskrrTune-method (as_tuned), 5</td>
</tr>
<tr>
<td></td>
<td>as_tuned, 5</td>
</tr>
<tr>
<td></td>
<td>as_tuned,tskrrHeterogeneous-method (as_tuned), 5</td>
</tr>
<tr>
<td></td>
<td>as_tuned,tskrrHomogeneous-method (as_tuned), 5</td>
</tr>
<tr>
<td></td>
<td>colMeans,linearFilter-method (alpha), 4</td>
</tr>
<tr>
<td></td>
<td>colnames,tskrr-method (labels.tskrr), 18</td>
</tr>
<tr>
<td></td>
<td>contour, 32, 33</td>
</tr>
<tr>
<td></td>
<td>create_grid, 6, 50</td>
</tr>
<tr>
<td></td>
<td>dim,tskrr-method, 7</td>
</tr>
<tr>
<td></td>
<td>dim.tskrr (dim,tskrr-method), 7</td>
</tr>
<tr>
<td></td>
<td>dimnames,tskrr-method (labels.tskrr), 18</td>
</tr>
<tr>
<td></td>
<td>drugSim (drugTargetInteraction), 7</td>
</tr>
<tr>
<td></td>
<td>drugtarget (drugTargetInteraction), 7</td>
</tr>
<tr>
<td></td>
<td>fitted, 9, 34, 42</td>
</tr>
<tr>
<td></td>
<td>fitted,linearFilter-method (fitted.tskrr), 9</td>
</tr>
<tr>
<td></td>
<td>fitted,tskrr-method (fitted.tskrr), 9</td>
</tr>
<tr>
<td></td>
<td>fitted.linearFilter (fitted.tskrr), 9</td>
</tr>
<tr>
<td></td>
<td>fitted.tskrr, 9</td>
</tr>
<tr>
<td></td>
<td>get_eigen, 42</td>
</tr>
<tr>
<td></td>
<td>get_eigen (response,tskrr-method), 38</td>
</tr>
<tr>
<td></td>
<td>get_grid (is_tuned), 17</td>
</tr>
<tr>
<td></td>
<td>get_kernel (response,tskrr-method), 38</td>
</tr>
<tr>
<td></td>
<td>get_kernelmatrix (response,tskrr-method), 38</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun, 10, 28, 43, 50, 51</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun,character-method (get_loo_fun), 10</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun,linearFilter-method (get_loo_fun), 10</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun,tskrrHeterogeneous-method (get_loo_fun), 10</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun,tskrrHomogeneous-method (get_loo_fun), 10</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun_tskrrHeterogeneous-method (get_loo_fun), 10</td>
</tr>
<tr>
<td></td>
<td>get_loo_fun,tskrrTune-method (get_loo_fun), 10</td>
</tr>
<tr>
<td></td>
<td>get_loss_values (is_tuned), 17</td>
</tr>
<tr>
<td></td>
<td>getter fuctions for linearFilter, 20</td>
</tr>
<tr>
<td></td>
<td>getters_linearFilter (alpha), 4</td>
</tr>
<tr>
<td></td>
<td>has_hat (response,tskrr-method), 38</td>
</tr>
<tr>
<td></td>
<td>has_imputed_values, 12</td>
</tr>
<tr>
<td></td>
<td>has_onedim (is_tuned), 17</td>
</tr>
<tr>
<td></td>
<td>hat, 13, 13</td>
</tr>
<tr>
<td></td>
<td>hat,tskrrHeterogeneous-method (hat), 13</td>
</tr>
<tr>
<td></td>
<td>hat,tskrrHomogeneous-method (hat), 13</td>
</tr>
<tr>
<td></td>
<td>hclust, 31</td>
</tr>
<tr>
<td></td>
<td>heatmap, 30</td>
</tr>
<tr>
<td></td>
<td>heterogeneous model, 39</td>
</tr>
<tr>
<td></td>
<td>image, 32</td>
</tr>
<tr>
<td></td>
<td>impute_tskrr, 13, 15, 16, 46</td>
</tr>
<tr>
<td></td>
<td>impute_tskrr.fit, 15</td>
</tr>
</tbody>
</table>
is_heterogeneous
  (response, tskrr-method), 38
is_homogeneous (response, tskrr-method), 38
is_imputed (has_imputed_values), 12
is_square (valid_dimensions), 52
is_symmetric, 16
is_tskrr (response, tskrr-method), 38
is_tuned, 17
isSymmetric, 16

Kmat_y2h_sc (proteinInteraction), 36

labels, tskrr-method (labels.tskrr), 18
labels.tskrr, 18
lambda (response, tskrr-method), 38
lambda.tskrr, 38
lambda, tskrrHeterogeneous-method (response, tskrr-method), 38
lambda, tskrrHomogeneous-method (response, tskrr-method), 38
linear_filter, 19, 20, 20
linearFilter, 4, 11, 20, 21, 23
linearFilter (linearFilter-class), 19
linearFilter-class, 19
loo, 11, 21, 23, 24, 28, 29, 31, 51
loo.linearFilter-method (loo), 21
loo, tskrrHeterogeneous-method (loo), 21
loo, tskrrHomogeneous-method (loo), 21
loo.b (loo_internal), 22
loo.c (loo_internal), 22
loo.e.skew (loo_internal), 22
loo.e.sym (loo_internal), 22
loo.e0.skew (loo_internal), 22
loo.e0.sym (loo_internal), 22
loo.i (loo_internal), 22
loo.j0 (loo_internal), 22
loo.r (loo_internal), 22
loo.v (loo_internal), 22
loo_internal, 11, 22, 51
loss, 23, 30
loss, permtest-method (loss), 23
loss, tskrr-method (loss), 23
loss, tskrrTune-method (loss), 23
loss.auc (loss_functions), 25
loss_functions, 23, 24, 25, 28, 51
loss.mse (loss_functions), 25

match_labels, 26
mean.linearFilter-method (alpha), 4

mean, linearFilter-method (alpha), 4
na_removed (alpha), 4
na_removed, linearFilter-method (alpha), 4
par, 33
permtest, 23, 27, 28, 29
permtest, tskrrHeterogeneous-method (permtest), 27
permtest, tskrrHomogeneous-method (permtest), 27
permtest, tskrrTune-method (permtest), 27
permtest-class, 28
permutations, 29
plot, 33
plot.tskrr, 30
plot_grid, 32
points, 33
predict, tskrr-method (predict.tskrr), 33
predict.tskrr, 33
print.permtest (permtest), 27
proteinInteraction, 36
residuals, 37
residuals, tskrr-method (residuals), 37
residuals.tskrr (residuals), 37
response, 42
response (response, tskrr-method), 38
response, tskrr-method, 38
row+colnames, 19
rowMeans, linearFilter-method (alpha), 4
rownames, tskrr-method (labels.tskrr), 18

symmetry (response, tskrr-method), 38
t.test, 29
targetSim (drugTargetInteraction), 7
test_symmetry, 40
tskrr, 5–7, 9, 11, 12, 15, 18, 19, 21, 24, 30, 32, 34, 35, 37–41, 41, 42–45, 47–49, 51, 52, 54
tskrr-class, 42
tskrr.fit, 43
tskrrHeterogeneous, 6, 42, 46, 48
tskrrHeterogeneous
  (tskrrHeterogeneous-class), 44
tskrrHeterogeneous-class, 44
tskrrHomogeneous, 6, 40, 42, 46, 48
tskrrHomogeneous
  (tskrrHomogeneous-class), 45
  tskrrHomogeneous-class, 45
  tskrrImppute, 12, 46
  tskrrImppute (tskrrImppute-class), 45
  tskrrImppute-class, 45
  tskrrImpputeHeterogeneous, 14
  tskrrImpputeHeterogeneous
    (tskrrImpputeHeterogeneous-class),
    46
  tskrrImpputeHeterogeneous-class, 46
  tskrrImpputeHomogeneous, 14
  tskrrImpputeHomogeneous
    (tskrrImpputeHomogeneous-class),
    46
  tskrrImpputeHomogeneous-class, 46
  tskrrTune, 5, 6, 17, 23, 32, 33, 35, 48, 50
  tskrrTune (tskrrTune-class), 47
  tskrrTune-class, 47
  tskrrTuneHeterogeneous
    (tskrrTuneHeterogeneous-class),
    48
  tskrrTuneHeterogeneous-class, 48
  tskrrTuneHomogeneous, 47
  tskrrTuneHomogeneous
    (tskrrTuneHomogeneous-class),
    48
  tskrrTuneHomogeneous-class, 48
  tune, 6, 24–26, 28, 32, 47, 48
  tune, matrix-method (tune), 48
  tune, tskrrHeterogeneous-method (tune), 48
  tune, tskrrHomogeneous-method (tune), 48
  update, 51
  update, tskrrHeterogeneous-method
    (update), 51
  update, tskrrHomogeneous-method
    (update), 51
valid_dimensions, 52
valid_labels, 53
weights, 55
weights
  (weights, tskrrHeterogeneous-method),
  54
weights, tskrrHeterogeneous-method, 54
weights, tskrrHeterogeneous-method
  (weights, tskrrHeterogeneous-method),
  54
which_imputed (has_imputed_values), 12
xnet (xnet-package), 3
xnet-package, 3