## Package 'maxstablePCA'

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**Title** Apply a PCA Like Procedure Suited for Multivariate Extreme Value Distributions

Type Package

**Description** Dimension reduction for multivariate data of extreme events with a PCA like procedure as described in Reinbott, Janßen, (2024), <doi:10.48550/arXiv.2408.10650>. Tools for necessary transformations of the data are provided.

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Imports nloptr

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#### compress

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compress

#### Description

Turn the given data into a compressed latent representation given by the fit of the max\_stable\_prcomp function. This is done by taking the max-matrix product of the data and the encoder matrix from the fit.

#### Usage

compress(fit, data)

#### Arguments

fit	max_stable_prcomp object. Data should be assumed to follow the same distri-
	bution as the data used in max_stable_prcomp.
data	array with same number of columns as the data of the fit object.

#### Value

An array of shape nrow(data), p giving the encoded representation of the data in p components which are also unit Frechet distributed which is to be takin into consideration for further analysis.

#### See Also

max\_stable\_prcomp(), maxmatmul()

#### Examples

```
# generate some data with the desired margins
dat <- matrix(evd::rfrechet(300), 100, 3)
maxPCA <- max_stable_prcomp(dat, 2)</pre>
```

```
# look at summary to obtain further information about
# loadings the space spanned and loss function
summary(maxPCA)
```

```
# transfrom data to compressed representation
# for a representation that is p-dimensional,
# preserves the max-stable structure and is numeric solution to
# optimal reconstruction.
compr <- compress(maxPCA, dat)</pre>
```

# For visual examination reconstruct original vector from compressed representation rec <- reconstruct(maxPCA, dat)</pre>

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elbe

A dataset about daily average river discharges (in  $m^3 / s$ ) for the Elbe river network at different measurement stations in Germany

#### Description

Measurements and geographical information about daily average river discharges in (m<sup>3</sup>/s) at 13 measurement stations from the Elbe river network from 31.12.1988 to 30.12.2010 for the train data and from 01.01.2010 to 31.12.2020 for the test data.

#### Usage

data(elbe)

#### Format

A named list containing differnent data files

- **train** A list containing the date of the measurement and measurements of the raw discharge data as data.frame at the 13 stations, and a data.frame containing the maximal discharge between the date "from" and "to". The blockmax dataset only considers the maximal value for the summer months June to September to reduce seasonal trends and temporal dependence.
- **test** Same structure as the two train data.frame objects, but only contains data from 01.01.2011 to 31.12.2020.
- **info** A data.frame object containing the station name, approximate latitude and longitude of the measurement station, the river measured and the next downstream station

#### Source

Datenportal der FGG Elbe https://www.elbe-datenportal.de

maxmatmul

Multiply two matrices with a matrix product that uses maxima instead of addition

#### Description

By calculating the entries with

$$(A \diamond B)_{ij} = \max_{i=1,\dots,l} A_{il} B_{lj}$$

for appropriate dimensions. Note that this operation is particularly useful when working with multivariate exreme value distributions, because, if the margins are standardized to standard Fréchet margins, then the max-matrix product of a matrix A and a multivariate extreme value distribution Z with standard Fréchet margins has the same margins up to scaling.

#### Usage

maxmatmul(A, B)

#### Arguments

A	a non-negative array of dim n, k
В	a non-negative array of dim k, l

#### Value

A non netgative array of dim n, l. The entries are given by the maximum of componentwise multiplication of rows from A and columns from B.

#### Examples

```
# Set up example matrices
A <- matrix(c(1,2,3,4,5,6), 2, 3)
B <- matrix(c(1,2,1,2,1,2), 3, 2)
# calling the function
m1 <- maxmatmul(A, B)
# can be used for matrix-vector multiplication as well
v <- c(7,4,7)
m2 <- maxmatmul(A, v)
m3 <- maxmatmul(v,v)</pre>
```

max\_stable\_prcomp Calculate max-stable PCA with dimension p for given dataset

#### Description

Find an optimal encoding of data of extremes using max-linear combinations by a distance minimization approach. Can be used to check if the data follows approximately a generalized max-linear model. For details on the statistical procedure it is advised to consult the articles "F. Reinbott, A. Janßen, Principal component analysis for max-stable distributions (https://arxiv.org/abs/2408.10650)" and "M.Schlather F. Reinbott, A semi-group approach to Principal Component Analysis (https://arxiv.org/abs/2112.04026)".

#### Usage

```
max_stable_prcomp(
   data,
   p,
   s = 3,
   n_initial_guesses = 150,
   norm = "11",
   optim_style = "full",
   ...
)
```

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#### Arguments

data	array or data.frame of n observations of d variables with unit Frechet margins. The max-stable PCA is fitted to reconstruct this dataset with a rank p approxi- mation.	
р	integer between 1 and ncol(data). Determines the dimension of the encoded state, i.e. the number of max-linear combinations in the compressed representation.	
S	(default = 3), numeric greater than 0. Hyperparameter for the	
n_initial_guesses		
	number of guesses to choose a valid initial value for optimization from. This procedure uses a pseudo random number generator so setting a seed is necessary for reproducibility. stable tail dependence estimator used in the calculation.	
norm	(delfault "l1") which norm to use for the spectral measure estimator, currently only l1 and sup norm "linfty" are available.	
optim_style	(delfault "full") choose between two different optimization strategies. The de- fault being "full" that optimizes both matrices simultaneously. the other choice "alternating" fixes one matrix then optimizes the other matrix until converged, then optimizes the other matrix in the same style. This can lead to more accurate results in some cases.	
	additional parameters passed to link{nloptr::slsqp()}	

#### Value

object of class max\_stable\_prcomp with slots p, inserted value of dimension, decoder\_matrix, an array of shape (d,p), where the columns represent the basis of the max-linear space for the reconstruction. encoder\_matrix, an array of shape (p,d), where the rows represent the loadings as max-linear combinations for the compressed representation. reconstr\_matrix, an array of shape (d,d), where the matrix is the mapping of the data to the reconstruction used for the distance minimization. loss\_fctn\_value, float representing the final loss function value of the fit. optim\_conv\_status, integer indicating the convergence of the optimizer if greater than 0.

#### Examples

```
# generate some data with the desired margins
dat <- matrix(evd::rfrechet(300), 100, 3)
maxPCA <- max_stable_prcomp(dat, 2)
# look at summary to obtain further information about
# loadings the space spanned and loss function
summary(maxPCA)
# transfrom data to compressed representation
# for a representation that is p-dimensional,
# preserves the max-stable structure and is numeric solution to
```

```
# optimal reconstruction.
```

```
compr <- compress(maxPCA, dat)</pre>
```

# For visual examination reconstruct original vector from compressed representation rec <- reconstruct(maxPCA, dat)</pre> reconstruct

#### Description

Map the data to the reconstruction given by the fit of the max\_stable\_prcomp function. This is done by taking the max-matrix product of the data and the reconstruction matrix from the fit.

#### Usage

```
reconstruct(fit, data)
```

#### Arguments

fit	max_stable_prcomp object. Data should be assumed to follow the same distribution as the data used in max_stable_prcomp.
data	array with same number of columns as the data of the fit object.

#### Value

An array of shape nrow(data), p giving the encoded representation of the data in p components which are also unit Frechet distributed which is to be takin into consideration for further analysis.

#### See Also

max\_stable\_prcomp(), maxmatmul()

#### Examples

```
# generate some data with the desired margins
dat <- matrix(evd::rfrechet(300), 100, 3)
maxPCA <- max_stable_prcomp(dat, 2)</pre>
```

```
# look at summary to obtain further information about
# loadings the space spanned and loss function
summary(maxPCA)
```

```
# transfrom data to compressed representation
# for a representation that is p-dimensional,
# preserves the max-stable structure and is numeric solution to
# optimal reconstruction.
compr <- compress(maxPCA, dat)</pre>
```

```
# For visual examination reconstruct original vector from compressed representation
rec <- reconstruct(maxPCA, dat)</pre>
```

summary.max\_stable\_prcomp

Print summary of a max\_stable\_prcomp object.

#### Description

Print summary of a max\_stable\_prcomp object.

#### Usage

```
## S3 method for class 'max_stable_prcomp'
summary(object, ...)
```

#### Arguments

object	max_stable_prcomp object. Data should be assumed to follow the same distri-
	bution as the data used in max_stable_prcomp.
	additional unused arguments.

#### Value

Same as base::print().

#### See Also

max\_stable\_prcomp()

transform\_orig\_margins

Transform the columns of a transformed dataset to original margins

#### Description

Since the dataset is intended to be transformed for PCA, this function takes a dataset transformed\_data and transforms the margins to the marginal distribution of the dataset orig\_data.

#### Usage

transform\_orig\_margins(transformed\_data, orig\_data)

#### Arguments

transformed\_data arraylike data of dimension n, d orig\_data arraylike data of dimension n, d

#### Value

array of dimension n,d with transformed columns of transformed\_data that follow approximately the same marginal distribution of orig\_data.

#### See Also

max\_stable\_prcomp(), transform\_unitfrechet(), [mev::fit.gev())] for information about why
to transform data

[mev::fit.gev())]: R:mev::fit.gev())

#### Examples

```
# create a sample
dat <- rnorm(1000)
transformed_dat <- transform_unitpareto(dat)</pre>
```

transform\_unitfrechet Transform the columns of a dataset to (approximately) unit Frechet margins

#### Description

Transforms columns of dataset to unit Frechet margins, to ensure the theoretical requirements are satisfied for the application of max\_stable\_prcomp using the empirical distribution function.

#### Usage

```
transform_unitfrechet(data)
```

#### Arguments

data array or vector with the data which columns are to be transformed

#### Value

array or vector of same shape and type as data with the transformed data with unit Frechet margins-

#### See Also

max\_stable\_prcomp(), transform\_orig\_margins(), [mev::fit.gev())] for information about why
to transform data.

[mev::fit.gev())]: R:mev::fit.gev())

#### transform\_unitpareto

#### Examples

```
# sample some data
dat <- rnorm(1000)
transformed_dat <- transform_unitfrechet(dat)
# Look at a plot of distribution
boxplot(transformed_dat)</pre>
```

plot(stats::ecdf(transformed\_dat))

transform\_unitpareto Transform the columns of a dataset to unit Pareto

#### Description

Transforms columns of dataset to unit Pareto margins, to ensure the theoretical requirements are satisfied for the application of max\_stable\_prcomp using the empirical distribution function.

#### Usage

```
transform_unitpareto(data)
```

#### Arguments

data array or vector with the data which columns are to be transformed

#### Value

array or vector of same shape and type as data with the transformed data with unit Frechet margins-

#### See Also

max\_stable\_prcomp(), transform\_orig\_margins(), [mev::fit.gev())] for information about why
to transform data.

[mev::fit.gev())]: R:mev::fit.gev())

#### Examples

```
# sample some data
dat <- rnorm(1000)
transformed_dat <- transform_unitfrechet(dat)</pre>
```

# Look at a plot of distribution boxplot(transformed\_dat) plot(stats::ecdf(transformed\_dat))

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