# Package 'maxstablePCA'

# September 6, 2024

<b>Title</b> Apply a PCA Like Procedure Suited for Multivariate Extreme Value Distributions
Type Package
<b>Description</b> 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.</doi:10.48550>
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Maintainer Felix Reinbott <felix.reinbott@ovgu.de></felix.reinbott@ovgu.de>
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Author Felix Reinbott [aut, cre]
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compress

Transform data to compact representation given by max-stable PCA

# **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)

# 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)

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

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maxmatmul

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

# Description

By calculating the entries with

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

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
B 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

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", ...)
```

# **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 approximation.
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 stable tail dependence estimator used in the calculation.
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.
norm	(delfault "11") which norm to use for the spectral measure estimator, currently only $11$ and sup norm "linfty" are available.
	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.

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# **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)

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

reconstruct

Obtain reconstructed data for PCA

# **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 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)

# 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)

# 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 distribution 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.

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# 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())
```

# **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))
```

# **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-

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# 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)
# Look at a plot of distribution
boxplot(transformed_dat)
plot(stats::ecdf(transformed_dat))</pre>
```

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