Package 'remote'

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Title Empirical Orthogonal Teleconnections in R

Version 1.2.3

Maintainer Tim Appelhans <tim.appelhans@gmail.com>

Description Empirical orthogonal teleconnections in R.

'remote' is short for 'R(-based) EMpirical Orthogonal TEleconnections'. It implements a collection of functions to facilitate empirical orthogonal teleconnection analysis. Empirical Orthogonal Teleconnections (EOTs) denote a regression based approach to decompose spatio-temporal fields into a set of independent orthogonal patterns. They are quite similar to Empirical Orthogonal Functions (EOFs) with EOTs producing less abstract results. In contrast to EOFs, which are orthogonal in both space and time, EOT analysis produces patterns that are orthogonal in either space or time.

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Depends R (>= 2.10), Rcpp (>= 0.10.3), raster, methods

Imports grDevices, gridExtra, latticeExtra, mapdata, scales, stats, utils

Suggests maps, lattice, grid, sp

LinkingTo Rcpp

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Author Tim Appelhans [cre, aut], Florian Detsch [aut], Thomas Nauss [ctb]

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remote-package

R EMpirical Orthogonal TEleconnections

Description

R EMpirical Orthogonal TEleconnections

Details

A collection of functions to facilitate empirical orthogonal teleconnection analysis. Some handy functions for preprocessing, such as deseasoning, denoising, lagging are readily available for ease of usage.

Author(s)

Tim Appelhans, Florian Detsch, Thomas Nauss

Maintainer: Tim Appelhans <tim.appelhans@gmail.com>

anomalize

References

Empirical Orthogonal Teleconnections H. M. van den Dool, S. Saha, A. Johansson (2000) Journal of Climate, Volume 13, Issue 8 (April 2000) pp. 1421 - 1435

Empirical methods in short-term climate prediction H. M. van den Dool (2007) Oxford University Press, Oxford, New York (2007)

See Also

remote is built upon Raster* classes from the raster::raster-package. Please see their documentation for data preparation etc.

anomalize

Create an anomaly RasterStack

Description

The function creates an anomaly RasterStack either based on the overall mean of the original stack, or a supplied reference RasterLayer. For the creation of seasonal anomalies use deseason().

Usage

```
anomalize(x, reference = NULL, ...)
```

Arguments

Х	a RasterStack
reference	an optional RasterLayer to be used as the reference
	additional arguments passed to raster::calc() (and, in turn, raster::writeRaster()) which is used under the hood

Value

an anomaly RasterStack

See Also

deseason(), denoise(), raster::calc()

Examples

```
data(australiaGPCP)
aus_anom <- anomalize(australiaGPCP)
opar <- par(mfrow = c(1,2))
plot(australiaGPCP[[10]], main = "original")
plot(aus_anom[[10]], main = "anomalized")
par(opar)</pre>
```

australiaGPCP

Monthly GPCP precipitation data for Australia

Description

Monthly Gridded Precipitation Climatology Project precipitation data for Australia from 1982/01 to 2010/12

Format

a RasterBrick with the following attributes

dimensions : 12, 20, 240, 348 (nrow, ncol, ncell, nlayers) resolution : 2.5, 2.5 (x, y) extent : 110, 160, -40, -10 (xmin, xmax, ymin, ymax) coord. ref. : +proj=longlat +ellps=WGS84 +towgs84=0,0,0,0,0,0,0 +no_defs

Details

Monthly Gridded Precipitation Climatology Project precipitation data for Australia from 1982/01 to 2010/12

References

The Version-2 Global Precipitation Climatology Project (GPCP) Monthly Precipitation Analysis (1979 - Present) Adler et al. (2003) Journal of Hydrometeorology, Volume 4, Issue 6, pp. 1147 - 1167 doi:10.1175/15257541(2003)004<1147:TVGPCP>2.0.CO;2

calcVar

Description

The function calculates the (optionally standardised) space-time variance of a RasterStack or Raster-Brick.

Usage

```
calcVar(x, standardised = FALSE, ...)
```

Arguments

Х	a RasterStack or RasterBrick
standardised	logical.
	currently not used

Value

the mean (optionally standardised) space-time variance.

Examples

data("pacificSST")

calcVar(pacificSST)

covWeight

Create a weighted covariance matrix

Description

Create a weighted covariance matrix

Usage

covWeight(m, weights, ...)

Arguments

m	a matrix (e.g. as returned by raster::getValues())
weights	a numeric vector of weights. For lat/lon data this can be produced with $getWeights()$
	additional arguments passed to stats::cov.wt()

Value

see stats::cov.wt()

See Also

stats::cov.wt()

cutStack

Shorten a RasterStack

Description

The function cuts a specified number of layers off a RrasterStack in order to create lagged Raster-Stacks.

Usage

cutStack(x, tail = TRUE, n = NULL)

Arguments

х	a RasterStack
tail	logical. If TRUE the layers will be taken off the end of the stack. If FALSE layers will be taken off the beginning.
n	the number of layers to take away.

Value

a RasterStack shortened by n layers either from the beginning or the end, depending on the specification of tail

Examples

data(australiaGPCP)

```
# 6 layers from the beginning
cutStack(australiaGPCP, tail = FALSE, n = 6)
# 8 layers from the end
cutStack(australiaGPCP, tail = TRUE, n = 8)
```

deg2rad

Description

Convert degrees to radians

Usage

deg2rad(deg)

Arguments

deg

vector of degrees to be converted to radians

Examples

data(vdendool)

latitude in degrees
degrees <- coordinates(vdendool)[, 2]
head(degrees)</pre>

```
## latitude in radians
radians <- deg2rad(coordinates(vdendool)[, 2])
head(radians)</pre>
```

denoise

Noise filtering through principal components

Description

Filter noise from a RasterStack by decomposing into principal components and subsequent reconstruction using only a subset of components

```
denoise(
    x,
    k = NULL,
    expl.var = NULL,
    weighted = TRUE,
    use.cpp = TRUE,
    verbose = TRUE,
    ...
)
```

Arguments

x	RasterStack to be filtered
k	number of components to be kept for reconstruction (ignored if expl.var is supplied)
expl.var	minimum amount of variance to be kept after reconstruction (should be set to NULL or omitted if k is supplied)
weighted	logical. If TRUE the covariance matrix will be geographically weighted using the cosine of latitude during decomposition (only important for lat/lon data)
use.cpp	logical. Determines whether to use Rcpp functionality, defaults to TRUE.
verbose	logical. If TRUE some details about the calculation process will be output to the console
	additional arguments passed to stats::princomp()

Value

a denoised RasterStack

See Also

anomalize(), deseason()

Examples

```
data("vdendool")
vdd_dns <- denoise(vdendool, expl.var = 0.8)
opar <- par(mfrow = c(1,2))
plot(vdendool[[1]], main = "original")
plot(vdd_dns[[1]], main = "denoised")
par(opar)</pre>
```

deseason

Create seasonal anomalies

Description

The function calculates anomalies of a RasterStack by supplying a suitable seasonal window. E. g. to create monthly anomalies of a raster stack of 12 layers per year, use cycle.window = 12.

```
## S4 method for signature 'RasterStackBrick'
deseason(x, cycle.window = 12L, use.cpp = FALSE, filename = "", ...)
## S4 method for signature 'numeric'
deseason(x, cycle.window = 12L)
```

eot

Arguments

x	An Raster* object or, alternatively, a numeric time series.
cycle.window	integer, defaults to 12. The window for the creation of the anomalies.
use.cpp	logical, defaults to FALSE. Determines whether or not to use Rcpp functional- ity. Only applies if x is a Raster* object.
filename	character. Output filename (optional).
	Additional arguments passed on to raster::writeRaster(), only considered if filename is specified.

Value

If x is a Raster* object, a deseasoned RasterStack; else a deseasoned numeric vector.

See Also

anomalize(), denoise()

Examples

```
data("australiaGPCP")
```

```
aus_dsn <- deseason(australiaGPCP, 12)</pre>
```

```
opar <- par(mfrow = c(1,2))
plot(australiaGPCP[[1]], main = "original")
plot(aus_dsn[[1]], main = "deseasoned")
par(opar)</pre>
```

EOT analysis of a predictor and (optionally) a response RasterStack

Description

Calculate a given number of EOT modes either internally or between RasterStacks.

```
## S4 method for signature 'RasterStackBrick'
eot(
    x,
    y = NULL,
    n = 1,
    standardised = TRUE,
    write.out = FALSE,
    path.out = ".",
    prefix = "remote",
```

```
reduce.both = FALSE,
type = c("rsq", "ioa"),
verbose = TRUE,
...
```

Arguments

х	a Raster* object used as predictor
У	a Raster* object used as response. If y is NULL, x is used as y
n	the number of EOT modes to calculate
standardised	logical. If FALSE the calculated r-squared values will be multiplied by the variance
write.out	logical. If TRUE results will be written to disk using path.out
path.out	the file path for writing results if write.out is TRUE. Defaults to current working directory
prefix	optional prefix to be used for naming of results if write.out is TRUE
reduce.both	logical. If TRUE both x and y are reduced after each iteration. If FALSE only y is reduced
type	the type of the link function. Defaults to 'rsq' as in original proposed method from <i>van den Dool 2000</i> . If set to 'ioa' index of agreement is used instead
verbose	logical. If TRUE some details about the calculation process will be output to the
	console

Details

For a detailed description of the EOT algorithm and the mathematics behind it, see the References section. In brief, the algorithm works as follows: First, the temporal profiles of each pixel xp of the predictor domain are regressed against the profiles of all pixels xr in the response domain. The calculated coefficients of determination are summed up and the pixel with the highest sum is identified as the 'base point' of the first/leading mode. The temporal profile at this base point is the first/leading EOT. Then, the residuals from the regression are taken to be the basis for the calculation of the next EOT, thus ensuring orthogonality of the identified teleconnections. This procedure is repeated until a predefined amount of n EOTs is calculated. In general, **remote** implements a 'brute force' spatial data mining approach to identify locations of enhanced potential to explain spatio-temporal variability within the same or another geographic field.

Value

if n = 1 an *EotMode*, if n > 1 an *EotStack* of n *EotModes*. Each *EotMode* has the following components:

- mode the number of the identified mode (1 n)
- *eot* the EOT (time series) at the identified base point. Note, this is a simple numeric vector, not of class ts

- coords_bp the coordinates of the identified base point
- *cell_bp* the cell number of the indeified base point
- cum_exp_var the (cumulative) explained variance of the considered EOT
- *r_predictor* the *RasterLayer* of the correlation coefficients between the base point and each pixel of the predictor domain
- *rsq_predictor* as above but for the coefficient of determination
- rsq_sums_predictor as above but for the sums of coefficient of determination
- *int_predictor* the *RasterLayer* of the intercept of the regression equation for each pixel of the predictor domain
- *slp_predictor* same as above but for the slope of the regression equation for each pixel of the predictor domain
- *p_predictor* the *RasterLayer* of the significance (p-value) of the the regression equation for each pixel of the predictor domain
- resid_predictor the RasterBrick of the reduced data for the predictor domain

Apart from *rsq_sums_predictor*, all *_*predictor* fields are also returned for the *_*response* domain, even if predictor and response domain are equal. This is due to that fact, that if not both fields are reduced after the first EOT is found, these *RasterLayers* will differ.

References

Empirical Orthogonal Teleconnections

H. M. van den Dool, S. Saha, A. Johansson (2000) Journal of Climate, Volume 13, Issue 8, pp. 1421-1435 doi:10.1175/15200442(2000)013<1421:EOT>2.0.CO;2

Empirical Methods in Short-Term Climate Prediction H. M. van den Dool (2007) Oxford University Press, Oxford, New York doi:10.1093/oso/9780199202782.001.0001

Examples

```
### EXAMPLE I
### a single field
```

data(vdendool)

```
plot(nh_modes, y = 2, show.bp = TRUE)
```

EotCycle

Description

EotCycle() calculates a single EOT and is controlled by the main eot() function

Usage

```
EotCycle(
    x,
    y,
    n = 1,
    standardised,
    orig.var,
    write.out,
    path.out,
    prefix,
    type,
    verbose,
    ....
)
```

Arguments

x	a ratser stack used as predictor
У	a RasterStack used as response. If y is NULL, x is used as y
n	the number of EOT modes to calculate
standardised	logical. If FALSE the calculated r-squared values will be multiplied by the variance
orig.var	original variance of the response domain
write.out	logical. If TRUE results will be written to disk using path.out
path.out	the file path for writing results if write.out is TRUE. Defaults to current working directory
prefix	optional prefix to be used for naming of results if write.out is TRUE
type	the type of the link function. Defaults to 'rsq' as in original proposed method from <i>Dool2000</i> . If set to 'ioa' index of agreement is used instead
verbose	logical. If TRUE some details about the calculation process will be output to the console
	If write.out = TRUE, further arguments passed to writeEot().

EotMode-class

Class EotMode

Description

Class EotMode

Slots

mode the number of the identified mode

name the name of the mode

eot the EOT (time series) at the identified base point. Note, this is a simple numeric vector

- coords_bp the coordinates of the identified base point
- cell_bp the cell number of the indeified base point

cum_exp_var the cumulative explained variance of the considered EOT mode

- r_predictor the RasterLayer of the correlation coefficients between the base point and each pixel of the predictor domain
- rsq_predictor as above but for the coefficient of determination of the predictor domain
- rsq_sums_predictor as above but for the sums of coefficient of determination of the predictor domain
- int_predictor the RasterLayer of the intercept of the regression equation for each pixel of the predictor domain
- slp_predictor same as above but for the slope of the regression equation for each pixel of the
 predictor domain
- p_predictor the RasterLayer of the significance (p-value) of the the regression equation for each pixel of the predictor domain
- resid_predictor the RasterBrick of the reduced data for the predictor domain
- r_response the RasterLayer of the correlation coefficients between the base point and each pixel of the response domain
- rsq_response as above but for the coefficient of determination of the response domain
- int_response the RasterLayer of the intercept of the regression equation for each pixel of the response domain
- p_response same the RasterLayer of the significance (p-value) of the the regression equation for each pixel of the response domain
- resid_response the RasterBrick of the reduced data for the response domain

EotStack-class Class EotStack

Description

Class EotStack

Slots

modes a list containing the individual 'EotMode's of the 'EotStack' names the names of the modes

geoWeight

Geographic weighting

Description

The function performs geographic weighting of non-projected long/lat data. By default it uses the cosine of latitude to compensate for the area distortion, though the user can supply other functions via f.

Usage

geoWeight(x, f = function(x) cos(x), ...)

Arguments

х	a Raster* object
f	a function to be used to the weighting. Defaults to $\cos(x)$
	additional arguments to be passed to f

Value

a weighted Raster* object

Examples

```
data(vdendool)
wgtd <- geoWeight(vdendool)
opar <- par(mfrow = c(1,2))
plot(vdendool[[1]], main = "original")
plot(wgtd[[1]], main = "weighted")
par(opar)</pre>
```

getWeights

Description

Calculate weights using the cosine of latitude to compensate for area distortion of non-projected lat/lon data

Usage

```
getWeights(x, f = function(x) cos(x), ...)
```

Arguments

х	a Raster* object
f	a function to be used to the weighting. Defaults to $\cos(x)$
	additional arguments to be passed to f

Value

a numeric vector of weights

Examples

```
data("australiaGPCP")
wghts <- getWeights(australiaGPCP)
wghts_rst <- australiaGPCP[[1]]
wghts_rst[] <- wghts
opar <- par(mfrow = c(1,2))
</pre>
```

plot(australiaGPCP[[1]], main = "data")
plot(wghts_rst, main = "weights")
par(opar)

lagalize

```
Create lagged RasterStacks
```

Description

The function is used to produce two lagged RasterStacks. The second is cut from the beginning, the first from the tail to ensure equal output lengths (provided that input lengths were equal).

Usage

lagalize(x, y, lag = NULL, freq = 12, ...)

Arguments

Х	a RasterStack (to be cut from tail)
У	a RasterStack (to be cut from beginning)
lag	the desired lag (in the native frequency of the RasterStack)
freq	the frequency of the RasterStacks
	currently not used

Value

a list with the two RasterStacks lagged by lag

Examples

```
data(pacificSST)
data(australiaGPCP)
# lag GPCP by 4 months
lagged <- lagalize(pacificSST, australiaGPCP, lag = 4, freq = 12)
lagged[[1]][[1]] #check names to see date of layer
lagged[[2]][[1]] #check names to see date of layer</pre>
```

Description

Calculate long-term means from an input 'RasterStack' (or 'RasterBrick') object. Ideally, the number of input layers should be divisable by the supplied cycle.window. For instance, if x consists of monthly layers, cycle.window should be a multiple of 12.

Usage

```
longtermMeans(x, cycle.window = 12L)
```

Arguments

Х	A 'RasterStack' (or 'RasterBrick') object.
cycle.window	'integer'. See deseason().

Value

If cycle.window equals nlayers(x) (which obviously doesn't make much sense), a 'RasterLayer' object; else a 'RasterStack' object.

names

Author(s)

Florian Detsch

See Also

deseason().

Examples

data("australiaGPCP")

longtermMeans(australiaGPCP)

names

Names of Eot* objects

Description

Get or set names of Eot* objects

Usage

S4 method for signature 'EotStack'
names(x)

S4 replacement method for signature 'EotStack'
names(x) <- value</pre>

S4 method for signature 'EotMode'
names(x)

S4 replacement method for signature 'EotMode'
names(x) <- value</pre>

Arguments

х	a EotMode or EotStack
value	name to be assigned

Value

if x is a EotStack, the names of all mdoes, if x is a EotMode, the name the respective mode

nmodes

Examples

```
data(vdendool)
nh_modes <- eot(vdendool, n = 2)
## mode names
names(nh_modes)
names(nh_modes) <- c("vdendool1", "vdendool2")
names(nh_modes)
names(nh_modes[2]])</pre>
```

nmodes

Number of modes of an EotStack

Description

Number of modes of an EotStack

Usage

S4 method for signature 'EotStack'
nmodes(x)

Arguments

x an EotStack

Details

retrieves the number of modes of an EotStack

Value

integer

Examples

data(vdendool)

 $nh_modes <- eot(vdendool, n = 2)$

nmodes(nh_modes)

nXplain

Description

The function identifies the number of modes needed to explain a certain amount of variance within the response field.

Usage

S4 method for signature 'EotStack'
nXplain(x, var = 0.9)

Arguments

х	an <i>EotStack</i>
var	the minimum amount of variance to be explained by the modes

Value

an integer denoting the number of EOTs needed to explain var

Note

This is a post-hoc function. It needs an *EotStack* created as returned by eot(). Depending on the potency of the identified EOTs, it may be necessary to compute a high number of modes in order to be able to explain a large enough part of the variance.

Examples

```
data(vdendool)
```

How many modes are needed to explain 25% of variance? nXplain(nh_modes, 0.25) pacificSST

Description

Monthly NOAA sea surface temperatures for the tropical Pacific Ocean from 1982/01 to 2010/12

Format

a RasterBrick with the following attributes

dimensions : 30, 140, 4200, 348 (nrow, ncol, ncell, nlayers) resolution : 1, 1 (x, y) extent : 150, 290, -15, 15 (xmin, xmax, ymin, ymax) coord. ref. : +proj=longlat +ellps=WGS84 +towgs84=0,0,0,0,0,0,0 +no_defs

Details

Monthly NOAA sea surface temperatures for the tropical Pacific Ocean from 1982/01 to 2010/12

References

Daily High-Resolution-Blended Analyses for Sea Surface Temperature Reynolds et al. (2007) Journal of Climate, Volume 20, Issue 22, pp. 5473 - 5496 doi:10.1175/2007JCL11824.1

plot

Plot an Eot* object

Description

This is the standard plotting routine for the results of eot(). Three panels will be drawn i) the predictor domain, ii) the response domain, iii) the time series at the identified base point

```
## S4 method for signature 'EotMode,ANY'
plot(
    x,
    y,
    pred.prm = "rsq",
    resp.prm = "r",
    show.bp = FALSE,
    anomalies = TRUE,
```

```
add.map = TRUE,
  ts.vec = NULL,
 arrange = c("wide", "long"),
 clr = NULL,
 locations = FALSE,
  • • •
)
## S4 method for signature 'EotStack,ANY'
plot(
 х,
 у,
 pred.prm = "rsq",
 resp.prm = "r",
  show.bp = FALSE,
  anomalies = TRUE,
  add.map = TRUE,
  ts.vec = NULL,
 arrange = c("wide", "long"),
 clr = NULL,
 locations = FALSE,
  . . .
)
```

Arguments

x	either an object of EotMode or EotStack as returned by eot()
У	integer or character of the mode to be plotted (e.g. 2 or "mode_2")
pred.prm	the parameter of the predictor to be plotted. Can be any of "r", "rsq", "rsq.sums", "p", "int" or "slp"
resp.prm	the parameter of the response to be plotted. Can be any of "r", "rsq", "rsq.sums", "p", "int" or "slp"
show.bp	logical. If TRUE a grey circle will be drawn in the predictor image to indicate the location of the base point
anomalies	logical. If TRUE a reference line will be drawn a 0 in the EOT time series
add.map	logical. If TRUE country outlines will be added to the predictor and response images
ts.vec	an (optional) time series vector of the considered EOT calculation to be shown as the x-axis in the time series plot
arrange	whether the final plot should be arranged in "wide" or "long" format
clr	an (optional) color palette for displaying of the predictor and response fields
locations	logical. If x is an EotStack, set this to TRUE to produce a map showing the locations of all modes. Ignored if x is an EotMode
	further arguments to be passed to raster::spplot()

predict

Methods (by class)

• plot(x = EotStack, y = ANY): EotStack

Examples

```
data(vdendool)
## claculate 2 leading modes
nh_modes <- eot(x = vdendool, y = NULL, n = 2,
                standardised = FALSE,
                verbose = TRUE)
## default settings
plot(nh_modes, y = 1) # is equivalent to
## Not run:
plot(nh_modes[[1]])
plot(nh_modes, y = 2) # shows variance explained by mode 2 only
plot(nh_modes[[2]]) # shows cumulative variance explained by modes 1 & 2
## showing the loction of the mode
plot(nh_modes, y = 1, show.bp = TRUE)
## changing parameters
plot(nh_modes, y = 1, show.bp = TRUE,
    pred.prm = "r", resp.prm = "p")
## change plot arrangement
plot(nh_modes, y = 1, show.bp = TRUE, arrange = "long")
## plot locations of all base points
plot(nh_modes, locations = TRUE)
## End(Not run)
```

predict

EOT based spatial prediction

Description

Make spatial predictions using the fitted model returned by eot(). A (user-defined) set of n modes will be used to model the outcome using the identified link functions of the respective modes which are added together to produce the final prediction.

predict

Usage

```
## S4 method for signature 'EotStack'
predict(object, newdata, n = 1, cores = 1L, filename = "", ...)
## S4 method for signature 'EotMode'
predict(object, newdata, n = 1, cores = 1L, filename = "", ...)
```

Arguments

object	an Eot* object
newdata	the data to be used as predictor
n	the number of modes to be used for the prediction. See nXplain() for calculat- ing the number of modes based on their explanatory power.
cores	integer. Number of cores for parallel processing.
filename	character, output filenames (optional). If specified, this must be of the same length as nlayers(newdata).
	further arguments passed to raster::calc(), and hence, raster::writeRaster().

Value

a RasterStack of nlayers(newdata)

See Also

raster::calc(), raster::writeRaster().

Examples

not very useful, but highlights the workflow

```
data(pacificSST)
data(australiaGPCP)
```

```
## compare results
opar <- par(mfrow = c(1,2))
plot(australiaGPCP[[13]], main = "original", zlim = c(0, 10))
plot(pred[[3]], main = "predicted", zlim = c(0, 10))
par(opar)</pre>
```

readEot

Description

Read Eot* related files from disk, e.g. for further use with predict() or plot().

Usage

readEot(x, prefix = "remote", suffix = "grd")

Arguments

х	character, search path for Eot* related files passed to list.files().
prefix	character, see writeEot() for details. Should be the same as previously supplied to eot().
suffix	character, file extension depending on the output file type of locally stored Eot* files, see raster::writeRaster().

Value

An Eot* object.

Author(s)

Florian Detsch

See Also

eot(), writeEot(), raster::writeRaster().

Examples

subset

Description

Extract a set of modes from an EotStack

Usage

```
## S4 method for signature 'EotStack'
subset(x, subset, drop = FALSE, ...)
## S4 method for signature 'EotStack,ANY,ANY'
```

```
## 54 method for signature Eotstack, ANY, A
x[[i]]
```

Arguments

x	EotStack to be subset
subset	integer or character. The modes to ectract (either by integer or by their names)
drop	if TRUE a single mode will be returned as an EotMode
	currently not used
i	number of EotMode to be subset

Value

an Eot* object

Examples

data(vdendool)

```
subs <- subset(nh_modes, 2:3) # is the same as
subs <- nh_modes[[2:3]]</pre>
```

```
## effect of 'drop=FALSE' when selecting a single layer
subs <- subset(nh_modes, 2)
class(subs)
subs <- subset(nh_modes, 2, drop = TRUE)
class(subs)
```

vdendool

Description

NCEP/NCAR reanalysis data of mean seasonal (DJF) 700 mb geopotential heights from 1948 to 1998

Format

a RasterBrick with the following attributes

dimensions : 14, 36, 504, 50 (nrow, ncol, ncell, nlayers) resolution : 10, 4.931507 (x, y) extent : -180, 180, 20.9589, 90 (xmin, xmax, ymin, ymax) coord. ref. : +proj=longlat +datum=WGS84 +ellps=WGS84 +towgs84=0,0,0

Details

NCEP/NCAR reanalysis data of mean seasonal (DJF) 700 mb geopotential heights from 1948 to 1998

Source

https://psl.noaa.gov/data/gridded/data.ncep.reanalysis.derived.pressure.html Original Source: NOAA National Center for Environmental Prediction

References

The NCEP/NCAR 40-year reanalysis project Kalnay et al. (1996) Bulletin of the American Meteorological Society, Volume 77, Issue 3, pp 437 - 471 doi:10.1175/15200477(1996)077<0437:TNYRP>2.0.CO;2

writeEot

Write Eot* objects to disk

Description

Write Eot* objects to disk. This is merely a wrapper around raster::writeRaster() so see respective help section for details.

writeEot

Usage

```
## S4 method for signature 'EotMode'
writeEot(x, path.out = ".", prefix = "remote", overwrite = TRUE, ...)
## S4 method for signature 'EotStack'
writeEot(x, path.out = ".", prefix, ...)
```

Arguments

х	an Eot* object
path.out	the path to the folder to write the files to
prefix	a prefix to be added to the file names (see Details)
overwrite	<pre>see raster::writeRaster(). Defaults to TRUE in writeEot()</pre>
	further arguments passed to raster::writeRaster()

Details

writeEot() will write the results of either an EotMode or an EotStack to disk. For each mode the following files will be written:

- *pred_r* the *RasterLayer* of the correlation coefficients between the base point and each pixel of the predictor domain
- pred_rsq as above but for the coefficient of determination
- pred_rsq_sums as above but for the sums of coefficient of determination
- *pred_int* the *RasterLayer* of the intercept of the regression equation for each pixel of the predictor domain
- *pred_slp* same as above but for the slope of the regression equation for each pixel of the predictor domain
- *pred_p* the *RasterLayer* of the significance (p-value) of the the regression equation for each pixel of the predictor domain
- pred_resid the RasterBrick of the reduced data for the predictor domain

Apart from *pred_rsq_sums*, all these files are also created for the response domain as *resp_**. These will be pasted together with the prefix & the respective mode so that the file names will look like, e.g.:

prefix_mode_n_pred_r.grd

for the *RasterLayer* of the predictor correlation coefficient of mode n using the standard *raster* file type (.grd).

Methods (by class)

writeEot(EotStack): EotStack

See Also

raster::writeRaster()

writeEot

Examples

End(Not run)

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