Package ‘remote’

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Type Package

Title Empirical Orthogonal Teleconnections in R

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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

RoxygenNote 5.0.1

NeedsCompilation yes

Repository CRAN

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R topics documented:

`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

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

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-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

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

Value

an anomaly RasterStack

See Also

`deseason`, `denoise`, `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)
```

`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
http://dx.doi.org/10.1175/1525-7541(2003)004<1147:TVGPCP>2.0.CO;2

calcVar — *Calculate space-time variance of a RasterStack or RasterBrick*

Description

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

Usage

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

Arguments

`x`	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 `getValues`)
`weights`	a numeric vector of weights. For lat/lon data this can be produced with `getWeights`
`...`	additional arguments passed to `cov.wt`

Value

see cov.wt

See Also

cov.wt

---

cutStack *Shorten a RasterStack*

Description

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

Usage

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

Arguments

x	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 *Convert degrees to radians*

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)

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

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

Usage

```
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 `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)
```

`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`.

Usage

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

## S4 method for signature 'numeric'
deseason(x, cycle.window = 12L)
```

Arguments

x	An object of class 'RasterStack' (or 'RasterBrick') or, alternatively, a 'numeric' time series.
cycle.window	Integer. The window for the creation of the anomalies.
use.cpp	Logical. Determines whether or not to use **Rcpp** functionality, defaults to TRUE. Only applies if x is a 'RasterStack' (or 'RasterBrick') object.
filename	character. Output filename (optional).
...	Additional arguments passed on to writeRaster, only considered if filename is specified.

Value

If x is a 'RasterStack' (or 'RasterBrick') object, a deseasoned 'RasterStack'; else a deseasoned 'numeric' vector.

See Also

anomalize, denoise

Examples

```
data("australiaGPCP")

aus_dsn <- deseason(australiaGPCP, 12)

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

---

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

Description

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

Usage

```
## 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

`x`	a RasterStack used as predictor
`y`	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
`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 *van den Dool 2000.* 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
`...`	not used at the moment

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` *EotMode*s. 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
http://journals.ametsoc.org/doi/abs/10.1175/1520-0442%282000%29013%3C1421%3AEOT%3E2.0.CO%3B2

Empirical methods in short-term climate prediction
H. M. van den Dool (2007)
Oxford University Press, Oxford, New York
https://global.oup.com/academic/product/empirical-methods-in-short-term-climate-prediction-97801992
cc=de&lang=en&

Examples

```
### EXAMPLE I
### a single field
data(vdendool)

## claculate 2 leading modes
nh_modes <- eot(x = vdendool, y = NULL, n = 2,
                standardised = FALSE,
                verbose = TRUE)

plot(nh_modes, y = 1, show.bp = TRUE)
plot(nh_modes, y = 2, show.bp = TRUE)
```

`EotCycle` *Calculate a single EOT*

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
`y`	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 *Dool2000*. 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
`...`	not used at the moment

`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

`slp_response` as above but for the slope 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

`x`	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)
```

`getWeights` *Calculate weights from latitude*

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

`x`	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))
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

`x`	a RasterStack (to be cut from tail)
`y`	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
```

`longtermMeans` *Calculate long-term means from a 'RasterStack'*

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

`x`	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.

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

## S4 method for signature 'EotMode'
names(x)

## S4 replacement method for signature 'EotMode'
names(x) <- value
```

Arguments

x	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

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]])
```

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 *Number of EOTs needed for variance explanation*

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

`x`	an *EotStack*
`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)

nh_modes <- eot(x = vdendool, y = NULL, n = 3,
                standardised = FALSE,
                verbose = TRUE)

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

`pacificSST` *Monthly SSTs for the tropical Pacific Ocean*

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
http://dx.doi.org/10.1175/2007JCLI1824.1

`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

Usage

```
## 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(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, ...)
```

Arguments

`x`	either an object of EotMode or EotStack as returned by `eot`
`y`	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 `spplot`

Methods (by class)

- `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.

Usage

```
## S4 method for signature 'EotStack'
predict(object, newdata, n = 1, ...)

## S4 method for signature 'EotMode'
predict(object, newdata, n = 1, ...)
```

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 calculating the number of modes based on their explnatory power.
`...`	further arguments to be passed to calc

Value

a *RasterStack* of `nlayers(newdata)`

Methods (by class)

- `EotMode`: EotMode

Examples

```
### not very useful, but highlights the workflow
data(pacificSST)
data(australiaGPCP)

## train data using eot()
train <- eot(x = pacificSST[[1:10]],
             y = australiaGPCP[[1:10]],
             n = 1)

## predict using identified model
pred <- predict(train,
                newdata = pacificSST[[11:20]],
                n = 1)

## 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)
```

subset — *Subset modes in EotStacks*

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'
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)

nh_modes <- eot(x = vdendool, y = NULL, n = 3,
                standardised = FALSE,
                verbose = TRUE)

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

## 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` *Mean seasonal (DJF) 700 mb geopotential heights*

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

`http://www.esrl.noaa.gov/psd/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
`http://journals.ametsoc.org/doi/abs/10.1175/1520-0477(1996)077%3C0437%3ATNYRP%3E2.0.CO%3B2`

`writeEot` *Write Eot* objects to disk*

Description

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

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

`x`	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`	see writeRaster. Defaults to `TRUE` in `writeEot`
`...`	further arguments passed to 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)

- `EotStack`: EotStack

See Also

writeRaster

Examples

```
data(vdendool)

nh_modes <- eot(x = vdendool, y = NULL, n = 2,
                standardised = FALSE,
                verbose = TRUE)

## write the complete EotStack
writeEot(nh_modes, prefix = "vdendool")

## write only one EotMode
writeEot(nh_modes[[2]], prefix = "vdendool")
```

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