

Package ‘Rfssa’

July 29, 2025

Type Package

Title Functional Singular Spectrum Analysis

Version 3.2.0

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URL <https://github.com/haghbinh/Rfssa>

Description

Methods and tools for implementing functional singular spectrum analysis and related techniques.

License GPL-3

Encoding UTF-8

LazyData true

LazyLoad true

RoxygenNote 7.3.2

Imports Rcpp, fda, lattice, plotly, shiny, Rssa, ggplot2, tibble,
RSpectra, rainbow, ftsa, dplyr, markdown

LinkingTo Rcpp, RcppArmadillo, RcppEigen,

Suggests knitr

Depends R (>= 4.0.0)

Repository CRAN

NeedsCompilation yes

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Date/Publication 2025-07-29 07:50:02 UTC

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<i>*.funts</i>	<i>Scalar Multiplication of Functional Time Series (funts)</i>
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Description

Perform scalar multiplication of a Functional Time Series (funts) object by either another funts object or a scalar value.

Usage

```
## S3 method for class 'funts'
obj1 * obj2
```

Arguments

- obj1 an object of class [funts](#) or a scalar value.
- obj2 an object of class [funts](#) or a scalar value.

Details

This function allows element-wise multiplication of a Functional Time Series (funts) object by another funts object or scalar value.

Value

An object of class `funts` representing the result of scalar multiplication.

See Also

`funts`

+.funts

Addition of Functional Time Series

Description

A method for functional time series (`funts`) addition and funts-scalar addition.

Usage

```
## S3 method for class 'funts'  
obj1 + obj2 = NULL
```

Arguments

obj1 an object of class `funts` or numeric.
obj2 an object of class `funts` or numeric.

Value

an object of class `funts`.

See Also

`funts`

- .funts

Subtract 'funts' Objects or a 'funts' Object and a Scalar

Description

This function allows subtraction between two 'funts' (functional time series) objects or between a 'funts' object and a scalar. It returns the resulting difference.

Usage

```
## S3 method for class 'funts'
obj1 - obj2 = NULL
```

Arguments

obj1 an object of class 'funts', representing the first 'funts' object.
obj2 an object of class 'funts' or a scalar.

See Also

[funts](#)

Examples

```
## Not run:
data("Callcenter")
y <- Callcenter
print(1 - y)

## End(Not run)
```

as.funts

Convert Object to a funts

Description

This function allows you to convert various types of objects into a functional time series ([funts](#)) object.

Usage

```
as.funts(obj, basis = NULL)
```

Arguments

- obj** the object to be converted. It can be an object of class `fd` (functional data) of the package ‘`fda`’, `fts` (functional time series) of the package ‘`rainbow`’ types.
- basis** an optional argument specifying the basis to be used for the resulting `funts` object when converting from `fts` objects. If not provided, a B-spline basis will be created by default.

Value

An object of class `funts`.

Note

Only objects of class `fd` (functional data) and `fts` (functional time series) can be converted to a `funts` object. Other types will result in an error.

See Also

`funts`, `create.bspline.basis`

Examples

```
require(rainbow)
class(Australiasmoothfertility)
x_funts1 <- as.funts(Australiasmoothfertility)
plot(x_funts1, main = "Australians Fertility")

require(fda)
bs <- create.bspline.basis(rangeval = c(15, 49), nbasis = 13)
fd_obj <- smooth.basis(argvals = Australiasmoothfertility$x, Australiasmoothfertility$y, bs)$fd

x_funts <- as.funts(fd_obj)
plotly_funts(x_funts,
  main = "Australians Fertility",
  ylab = "Fertility rate",
  xlab = "Age"
)
```

Callcenter

Callcenter Dataset: Number of Calls for a Bank

Description

This dataset represents a small call center for an anonymous bank (Brown et al., 2005). It provides detailed information about the exact times of calls that were connected to the center throughout the year 1999, from January 1 to December 31.

Format

A functional time series object of class ‘funts’ with the following fields:

time the time index indicating when the calls occurred.

coefs the coefficients corresponding to the B-spline basis functions.

basisobj the basis functions used for the functional representation.

dimSupp the dimension support of the functional data.

Details

The data have been converted into a functional time series using a B-spline basis system with 22 basis functions. The resulting dataset is stored as a functional time series object of class ‘funts’. You can load the raw data using the function [loadCallcenterData](#). See [funts](#) for more details.

References

1. Brown, L., Gans, N., Mandelbaum, A., Sakov, A., Shen, H., Zeltyn, S., & Zhao, L. (2005). Statistical analysis of a telephone call center: A queueing-science perspective. *Journal of the American Statistical Association*, **100**(469), 36-50.

See Also

[loadCallcenterData](#), [funts](#)

Examples

```
## Not run:
# Load the Callcenter dataset
data("Callcenter")

## End(Not run)
```

eval.funts

Evaluate a Functional Time Series (funts) Object on a Given Grid

Description

This function allows you to evaluate a Functional Time Series (funts) object on a specified grid of argument values. The result is a list of matrices, each matrix corresponding to one dimension of the functional data.

Usage

```
eval.funts(argvals, obj)
```

Arguments

- `argvals` a list or numeric vector specifying the grid points at which to evaluate the functional time series. For multivariate functional data, provide a list of grids corresponding to each dimension.
- `obj` an object of class `funts` to be evaluated.

Details

The `argvals` argument can be a list of grids for multivariate functional data. The function handles both functional basis and empirical basis cases for evaluation. For empirical basis with irregular grids, a warning is issued as this feature is under development.

Value

A list of matrices, where each matrix represents the evaluated values of the functional data on the specified grid.

See Also

`funts`

Examples

```
data("Montana")
y <- Montana
u <- seq(0, 23, len = 4)
v <- seq(1, 33, len = 3)
grid <- list(u, list(v, v))
eval.funts(grid, y)
```

fforecast

Functional Singular Spectrum Analysis Recurrent and Vector Forecasting

Description

Perform functional singular spectrum analysis (FSSA) recurrent forecasting (FSSA R-forecasting) or vector forecasting (FSSA V-forecasting) on univariate or multivariate functional time series (`funts`) observed over a one-dimensional domain.

Usage

```
fforecast(
  U,
  groups = as.list(1L:10L),
  len = 1,
  method = "recurrent",
```

```

    only.new = TRUE,
    tol = NULL
  )

```

Arguments

<code>U</code>	an object of class <code>fssa</code> holding the decomposition.
<code>groups</code>	a list of numeric vectors where each vector includes indices of elementary components of a group used for reconstruction and forecasting.
<code>len</code>	integer, the desired length of the forecasted FTS.
<code>method</code>	a character string specifying the type of forecasting to perform: - "recurrent" for FSSA R-forecasting. - "vector" for FSSA V-forecasting.
<code>only.new</code>	logical, if 'TRUE' then only forecasted FTS are returned, whole FTS otherwise.
<code>tol</code>	a double specifying the tolerated error in the approximation of the matrix used in forecasting algorithms.

Value

An object of class 'fforecast' which is a list of objects of class `funts`, where each one corresponds to a forecasted group.

Examples

```

## Not run:
data("Callcenter")
U <- fssa(Callcenter, L = 28)
groups <- list(1, 1:7)

## Perform FSSA R-forecast
pr_R <- fforecast(
  U = U, groups = groups, only.new = FALSE,
  len = 30, method = "recurrent"
)

plot(pr_R, group_index = 1 )

plotly_funts(pr_R[[2]], main = "group = '1:7'")

## Perform FSSA V-forecast
pr_V <- fforecast(U = U, groups = groups, len= 30, method = "vector")

plot(pr_V, group_index = 1)

plotly_funts(pr_V[[2]], type = "3Dline" , main = "group = '1:7'")

# Multivariate forecasting example:
data("Montana")
time <- Montana$time
grid <- list(0:23, list(1:33, 1:33))

```



```

montana <- eval.funts(Montana, argvals = grid)
montana[[2]] <- array(
  scale(montana[[2]][, , ]),
  center = min(montana[[2]][, , ]),
  scale = max(montana[[2]][, , ]) - min(montana[[2]][, , ])
),
dim = c(33, 33, 133)
)
## Kernel density estimation of pixel intensity
NDVI <- matrix(NA, nrow = 512, ncol = 133)
for (i in 1:133) NDVI[, i] <- (density(montana[[2]][, , i], from = 0, to = 1)$y)

## Define functional objects
bs1 <- Montana$basis[[1]]

require(fda)
bs2 <- create.bspline.basis(nbasis = 15)
Y <- funts(X = list(montana[[1]], NDVI), basisobj = list(bs1, bs2),
  vnames = c("Temperature", "NDVI Density"),
  dnames = c("Time", "NDVI"),
  tname = "Date")

plotly_funts(Y,
  main = c("Temperature", "NDVI"),
  xticklocs = list(c(0, 6, 12, 18, 23), seq(1, 512, len = 9)),
  xticklabels = list(c(0, 6, 12, 18, 23), seq(0, 1, len = 9))
)

U <- fssa(Y = Y, L = 45)
plotly_funts(U$Lsingf[[1]])
plot(U$Lsingf[[2]])

groups <- list(1, 1:3)
pr_R <- fforecast(U = U, groups = groups,
  only.new = FALSE, len = 10, method = "recurrent")
plot(pr_R)
plotly_funts(pr_R[[2]], main = "Recurrent method, group = '1:3'")

pr_V <- fforecast(U = U, groups = groups, len = 10, method = "vector")
plot(pr_V, group_index = 1)
plotly_funts(pr_V[[2]], main = "Vector method, group = '1:3'")

## End(Not run)

```

Description

Calculate the bootstrap prediction interval for functional singular spectrum analysis (FSSA) forecasting predictions of univariate functional time series ([funts](#)) observed over a one-dimensional domain.

Usage

```
fpredinterval(
  Y,
  O = floor(Y$N * 0.7),
  L = floor((Y$N * 0.7)/12),
  ntriples = 10,
  Bt = 100,
  h = 1,
  alpha = 0.05,
  method = "recurrent",
  tol = 10^-3
)
```

Arguments

Y	an object of class funts .
O	a positive integer specifying the training set size.
L	a positive integer specifying the window length.
ntriples	the number of eigentriples to use for forecasts.
Bt	a positive integer specifying the number of bootstrap samples.
h	an integer specifying the forecast horizon.
alpha	a double ($0 < \alpha < 1$) specifying the significance level.
method	a character string: "recurrent" or "vector" forecasting.
tol	a double specifying tolerated error in the approximation.

Value

a list of numeric vectors: point forecast, lower, and upper bounds.

Examples

```
## Not run:
data("Callcenter")
pred_interval <- fpredinterval(
  Y = Callcenter, O = 310,
  L = 28, ntriples = 7, Bt = 10000, h = 3
)

# Plot the forecast and prediction interval using ggplot
df <- data.frame(
  x = 1:240,
```

```

    y = pred_interval$forecast,
    lower = pred_interval$lower,
    upper = pred_interval$upper
  )
  require(ggplot2)
  # Create the ggplot
  ggplot(df, aes(x = x, y = y)) +
    geom_line(linewidth = 1.2) +
    scale_x_continuous(
      name = "Time",
      breaks = c(1, 60, 120, 180, 240),
      labels = c("00:00", "06:00", "12:00", "18:00", "24:00"),
    ) +
    scale_y_continuous(name = "Sqrt of Call Numbers") +
    ggtitle("Prediction Intervals for Jan. 3, 2000") +
    geom_ribbon(aes(ymin = lower, ymax = upper), fill = "darkolivegreen3", alpha = 0.3) +
    theme_minimal()

## End(Not run)

```

freconstruct

Reconstruction Stage of Functional Singular Spectrum Analysis

Description

Reconstruct univariate or multivariate functional time series ([funts](#)) objects from functional singular spectrum analysis ([fssa](#)) objects, including Grouping and Hankelization steps. This function performs the reconstruction step for either univariate functional singular spectrum analysis (ufssa) or multivariate functional singular spectrum analysis (mfssa), depending on the input.

Usage

```
freconstruct(U, groups = as.list(1L:10L))
```

Arguments

U	an object of class fssa .
groups	a list of numeric vectors, each vector includes indices of elementary components of a group used for reconstruction.

Value

A named list of objects of class [funts](#) that are reconstructed according to the specified groups and a numeric vector of eigenvalues.

Note

Refer to [fssa](#) for an example on how to run this function starting from [fssa](#) objects.

See Also

[fssa](#), [funts](#)

Examples

```
data("Callcenter")
L <- 28
U <- fssa(Callcenter, L)

# FSSA Reconstruction step:
gr <- list(1, 2:3, 4:5, 6:7, 1:7)
Q <- freconstruct(U, gr)
plot(Q[[1]],
     main = "Call Center Mean Component")
plot(Q[[2]],
     main = "Call Center First Periodic Component")

#----- Multivariate FSSA Example on bivariate -----
## temperature curves and smoothed images of vegetation
## Not run:
data("Montana")
L <- 45
U <- fssa(Montana, L)

# MFSSA Reconstruction step:
Q <- freconstruct(U = U, groups = list(1, 2, 3))
plotly_funts(Q[[1]],
  main = c("Temperature Curves Mean", "NDVI Images Mean"),
  color_palette = "RdYlGn",
  xticklabels = list(
    c("00:00", "06:00", "12:00", "18:00", "24:00"),
    c("113.40\u00B0 W", "113.30\u00B0 W")
  ),
  xticklocs = list(c(1, 6, 12, 18, 24), c(1, 33)),
  yticklabels = list(NA, c("48.70\u00B0 N", "48.77\u00B0 N")),
  yticklocs = list(NA, c(1, 33))
) # mean

plotly_funts(Q[[2]],
  main = c("Temperature Curves Periodic", "NDVI Images Periodic"),
  color_palette = "RdYlGn",
  xticklabels = list(
    c("00:00", "06:00", "12:00", "18:00", "24:00"),
    c("113.40\u00B0 W", "113.30\u00B0 W")
  ),
  xticklocs = list(c(1, 6, 12, 18, 24), c(1, 33)),
  yticklabels = list(NA, c("48.70\u00B0 N", "48.77\u00B0 N")),
  yticklocs = list(NA, c(1, 33))
) # periodic

plot(Q[[3]],
  obs = 3,
```

```

    main = c("Temperature Curves Trend", "NDVI Images Trend,")
  ) # trend

## End(Not run)

```

fssa

*Functional Singular Spectrum Analysis (FSSA)***Description**

This function performs the decomposition (embedding and functional SVD steps) for univariate (ufssa) or multivariate (mfssa) functional singular spectrum analysis based on the input data type. The input can be a univariate or multivariate functional time series ([funts](#)) object.

Usage

```
fssa(Y, L = Y$N/2, ntriples = 20, type = "ufssa")
```

Arguments

Y	an object of class funts .
L	a positive integer, the window length, the default is half of FTS length.
ntriples	a positive integer, the number of eigentriples for the decomposition.
type	a string indicating the type of FSSA: "ufssa" (default for univariate FTS) or "mfssa" (default for multivariate FTS).

Value

An object of class fssa, containing functional objects, eigenvalues, window length, and original data.

Examples

```

data("Callcenter")

# FSSA Decomposition step:
L <- 28
U <- fssa(Callcenter, L)
plot(U, type = "values", d = 10)
plot(U, type = "vectors", d = 4)
plot(U, type = "paired", d = 6)
plot(U, type = "lcurves", d = 4, vars = 1)
plot(U, type = "lheats", d = 4)
plot(U, type = "wcor", d = 10)
plotly_funts(U$Lsingf[[1]])
plot(U$Lsingf[[2]])

```

```
## Not run:
#----- Multivariate FSSA Example on bivariate -----
## temperature curves and smoothed images of vegetation
data("Montana")

# MFSSA Decomposition step:
L <- 45
U <- fssa(Montana, L)
plot(U, type = "values", d = 10)
plot(U, type = "vectors", d = 4)
plot(U, type = "lheats", d = 4)
plot(U, type = "lcurves", d = 4, vars = 1)
plot(U, type = "paired", d = 6)
plot(U, type = "periodogram", d = 4)
plot(U, type = "wcor", d = 10)
plotly_funts(U$Lsingf[[1]])
plot(U$Lsingf[[2]])

## End(Not run)
```

funts

Functional Time Series (funts) Class

Description

The ‘funts’ class is designed to encapsulate functional time series objects, including both univariate (FTS) and multivariate (MFTS) forms. It provides a versatile framework for creating and manipulating ‘funts’ objects, accommodating various basis systems and dimensions.

Usage

```
funts(
  X,
  basisobj,
  argval = NULL,
  method = "data",
  start = 1,
  end = NULL,
  vnames = NULL,
  dnames = NULL,
  tname = NULL
)
```

Arguments

X A matrix, three-dimensional array, or a list of matrix or array objects. When ‘method="data"’, it represents the observed curve values at discrete sampling

points or argument values. When `'method="coefs"'`, `'X'` specifies the coefficients corresponding to the basis system defined in `'basisobj'`. If `'X'` is a list, it defines a multivariate FTS, with each element being a matrix or three-dimensional array object. In matrix objects, rows correspond to argument values, and columns correspond to the length of the FTS. In three-dimensional array objects, the first and second dimensions correspond to argument values, and the third dimension to the length of the FTS.

<code>basisobj</code>	An object of class <code>'basisfd'</code> , a matrix of empirical basis, or a list of <code>'basisfd'</code> or empirical basis objects. For empirical basis, rows correspond to basis functions, and columns correspond to grid points.
<code>argval</code>	A vector list of length <code>'p'</code> which is a set of argument values corresponding to the observations in <code>X</code> . Each entry in this list should either be a numeric value or a list of numeric elements, depending on the dimension of the domain the variable is observed over. It can even vary from one variable to another, If it be <code>NULL</code> , the default value for <code>argval</code> are the integers 1 to <code>n</code> , where <code>n</code> is the size of the first dimension in argument <code>X</code> .
<code>method</code>	Determines the type of the <code>'X'</code> matrix: <code>"coefs"</code> or <code>"data."</code>
<code>start</code>	The time of the first observation. It can be a single positive integer or an object of classes <code>'Date'</code> , <code>'POSIXct'</code> , or <code>'POSIXt'</code> , specifying a natural time unit.
<code>end</code>	The time of the last observation, specified in the same way as <code>'start'</code> .
<code>vnames</code>	a vector of strings specifies the variable names
<code>dnames</code>	list of vector of strings specifies the variable domain names
<code>tname</code>	a string specifies the time index name

Value

An instance of the `'funts'` class containing functional time series data.

An instance of the `'funts'` class containing functional time series data.

See Also

[fssa](#)

Examples

```
# 1D FTS example: Callcenter dataset
require(fda)
Call_data <- loadCallcenterData()
D <- matrix(sqrt(Call_data$calls), nrow = 240)
bs1 <- create.bspline.basis(c(0, 23), 22)
Y <- funts(X = D, basisobj = bs1, start = as.Date("1999-1-1"),
           vnames = "Sqrt of Call Numbers",
           dnames = "Time (6 minutes aggregated)",
           tname = "Date" )

plot(Y, lwd = 2, npts = 200, col = "deepskyblue4",
     main = "Call Center Data")
```

```

)

# -----2D Multivariate Example: Montana dataset
# Temperature curves and smoothed images of vegetation
montana_data <- loadMontanaData()
Temp <- montana_data$Temp
NDVI <- montana_data$NDVI
Montana_Data <- list(Temp / sd(Temp), NDVI)
bs1 <- create.bspline.basis(c(0, 23), 11)
bs2 <- create.bspline.basis(c(1, 33), 13)
bs2d <- list(bs2, bs2)
bsmv <- list(bs1, bs2d)
Y <- funts(X = Montana_Data, basisobj = bsmv,
           start = as.Date("2008-01-01"),
           end = as.Date("2013-09-30"),
           vnames = c("Normalized Temperature (\u00B0C)" , "NDVI"),
           dnames = list("Time", c("Latitude", "Longitude")),
           tname = "Date"
)

plot(Y, main = "Montana dataset")

```

is.funts

*Check if an object is of class 'funts'***Description**

Check if an object is of class 'funts'

Usage

```
is.funts(obj)
```

Arguments

obj The object to check.

Value

TRUE if the object is of class 'funts', FALSE otherwise.

Examples

```
data("Callcenter")
is.funts(Callcenter)
```

launchApp

Launch Shiny Application for FSSA Demonstration

Description

This function launches a Shiny app to facilitate the understanding of univariate or multivariate Functional Singular Spectrum Analysis ([fssa](#)). The app enables users to perform univariate or multivariate FSSA on various data types, including simulated and real data provided by the server. Users can also upload their own data. The app supports simultaneous comparisons of different methods, such as multivariate vs. univariate FSSA, and allows users to select the number and types of basis elements used for estimating Functional Time Series ([funts](#)) objects. It offers plotting capabilities for both [funts](#) and [fssa](#).

Usage

```
launchApp(type = "ufssa")
```

Arguments

type Type of FSSA with options of type = "ufssa" or type = "mfssa".

Value

A shiny application object.

Examples

```
## Not run:
launchApp()

## End(Not run)
```

length.funts

Length of Functional Time Series

Description

Returns the length of a "funts" object.

Usage

```
## S3 method for class 'funts'
length(x)
```

Arguments

`x` an object of class "funts" .

Examples

```
data("Callcenter")
length(Callcenter)
```

loadAustinData

Load Austin Temperature Data from GitHub Repository

Description

This function retrieves the Austin Temperature dataset from a GitHub repository hosted at <https://github.com/hagbhin/dataset>. Hosting datasets on GitHub rather than including them in the Rfssa R package conserves storage space. The Austin Temperature dataset contains intraday hourly temperature curves measured in degrees Celsius from January 1973 to July 2023, recorded once per month. The returned object is a raw dataset in 'list' format;

Usage

```
loadAustinData()
```

Format

Containing two matrix objects:

Austin Temperature Data A 24 by 607 matrix of discrete samplings of intraday hourly temperature curves, one day per month.

Utqiagvik Temperature Data A 24 by 607 matrix of discrete samplings of intraday hourly temperature curves, one day per month.

References

Trinka, J., Hagbhin, H., Shang, H., Maadooliat, M. (2023). Functional Time Series Forecasting: Functional Singular Spectrum Analysis Approaches. Stat, e621.

Examples

```
Austin_raw <- loadAustinData()
str(Austin_raw)
```

loadCallcenterData	<i>Load Callcenter Data from GitHub Repository</i>
--------------------	--

Description

This function retrieves the Callcenter dataset from the Rfssa_dataset repository on GitHub (<https://github.com/hagbinh/dataset>). The Callcenter dataset represents a small call center for an anonymous bank. It provides precise call timing data from January 1 to December 31, 1999. The data is aggregated into 6-minute intervals on each day. The returned object is a raw dataset in dataframe format; it is not a 'funts' class object. This raw data can then be further processed and converted into a 'funts' object named 'Callcenter'. See [funts](#) for more details on working with functional time series of class 'funts'.

Usage

```
loadCallcenterData()
```

Format

a dataframe with 87,600 rows and 5 variables:

calls number of calls in a 6-minute aggregated interval.

u numeric vector indicating the aggregated interval.

Date date and time of call count recording.

Day weekday associated with Date.

Month month associated with Date.

References

1. Brown, L., Gans, N., Mandelbaum, A., Sakov, A., Shen, H., Zeltyn, S., & Zhao, L. (2005). Statistical analysis of a telephone call center: A queueing-science perspective. *Journal of the American Statistical Association*, **100**(469), 36-50.
2. Shen, H., & Huang, J. Z. (2005). Analysis of call center arrival data using singular value decomposition. *Applied Stochastic Models in Business and Industry*, **21**(3), 251-263.
3. Huang, J. Z., Shen, H., & Buja, A. (2008). Functional principal components analysis via penalized rank one approximation. *Electronic Journal of Statistics*, **2**, 678-695.
4. Maadooliat, M., Huang, J. Z., & Hu, J. (2015). Integrating data transformation in principal components analysis. *Journal of Computational and Graphical Statistics*, **24**(1), 84-103.

See Also

[funts](#)

Examples

```
require(fda)
# Load Callcenter data
Call_data <- loadCallcenterData()
D <- matrix(sqrt(Call_data$calls), nrow = 240)

# Define basis functions
bs1 <- create.bspline.basis(c(0, 23), 22)

Y <- funts(X = D, basisobj = bs1)
```

loadJambiData

Load Jambi MODIS Data from GitHub Repository

Description

This function retrieves the ‘Jambi’ dataset from a GitHub repository hosted at https://github.com/hagbhin/dataset/Rfssa_data. The Jambi dataset contains normalized difference vegetation index (NDVI) and enhanced vegetation index (EVI) image data from NASA’s MODerate-resolution Imaging Spectroradiometer (MODIS) with global coverage at a 250 m² resolution. The dataset covers the Jambi Province, Indonesia, known for various forested land uses, including natural forests and plantations. Monitoring land cover changes is crucial, especially in the context of forest exploitation and conservation efforts. Seasonal variations significantly impact long-term land cover changes. Data collection began on February 18, 2000, and continued until July 28, 2019, with data recorded every 16 days. This dataset is valuable for studying vegetative land cover changes in the region. The returned object is a raw dataset in ‘list’ format;

Usage

```
loadJambiData()
```

Format

A list containing two arrays, each with dimensions 33 by 33 by 448. One array represents NDVI image data, and the other represents EVI image data. The list also contains a date vector of length 448, specifying the capture date for each 33 by 33 image.

Source

[MODIS Product Information](<https://lpdaac.usgs.gov/products/mod13q1v006/>)

References

1. Lambin, E., Geist, H., Lepers, E. (1999). Dynamics of Land-Use and Land-Cover Change in Tropical Regions. *Annual Review of Environment and Resources*, 205-244.

See Also

- The dataset object loaded by this function.

Examples

```
Jambi_raw <- loadJambiData()
str(Jambi_raw)
```

loadMontanaData

*Load Montana Data from GitHub Repository***Description**

This function retrieves the Montana dataset from a GitHub repository hosted at https://github.com/hagbhin/dataset/Rfssa_data. Hosting datasets on GitHub rather than including them in the Rfssa R package conserves storage space. The Montana dataset contains intraday hourly temperature curves measured in degrees Celsius and normalized difference vegetation index (NDVI) image data. Both types of data are recorded near Saint Mary, Montana, USA. The NDVI images cover a region located between longitudes of 113.30 degrees West and 113.56 degrees West and latitudes of 48.71 degrees North and 48.78 degrees North. For each recorded intraday temperature curve, an NDVI image was captured on the same day every 16 days, starting from January 1, 2008, and ending on September 30, 2013. The dataset is valuable for environmental analysis, especially in the context of studying the impact of temperature changes on vegetation. Combining both temperature and NDVI data can reveal more informative patterns and insights. The returned object is a raw dataset in 'list' format; This raw data can then be further processed and converted into a 'funts' object named 'Montana'. See [funts](#) for more details on working with functional time series of class 'funts'.

Usage

```
loadMontanaData()
```

Format

A list containing two components:

Temperature Data A 24 by 133 matrix of discrete samplings of intraday hourly temperature curves.

NDVI Images An array with dimensions 33 by 33 by 133, where each 33 by 33 slice represents an NDVI image.

References

1. Diamond, H. J., Karl, T., Palecki, M. A., Baker, C. B., Bell, J. E., Leeper, R. D., Easterling, D. R., Lawrimore, J. H., Meyers, T. P., Helfert, M. R., Goodge, G., and Thorne, P.W. (2013). U.S. climate reference network after one decade of operations: status and assessment. [Read More](<https://www.ncdc.noaa.gov/crn/qcdatasets.html>). Last accessed April 2020.
2. Tuck, S. L., Phillips, H. R., Hintzen, R. E., Scharlemann, J. P., Purvis, A., and Hudson, L. N. (2014). MODISTools – downloading and processing MODIS remotely sensed data in R. Ecology and Evolution, 4(24):4658–4668.

See Also[funts](#)**Examples**

```

require(fda)
# Load Montana data
montana_data <- loadMontanaData()

# Extract variables
Temp <- montana_data$Temp
NDVI <- montana_data$NDVI

# Create a list for Montana data
Montana_Data <- list(Temp / sd(Temp), NDVI)

# Define basis functions
bs1 <- create.bspline.basis(c(0, 23), 11)
bs2 <- create.bspline.basis(c(1, 33), 13)
bs2d <- list(bs2, bs2)
bsmv <- list(bs1, bs2d)

# Convert to funts object
Y <- funts(X = Montana_Data, basisobj = bsmv,
          start = as.Date("2008-01-01"),
          end = as.Date("2013-09-30"),
          vnames = c("Normalized Temperature (\u00B0C)" , "NDVI"),
          dnames = list("Time", c("Latitude", "Longitude")),
          tname = "Date"
)
```

loadUtqiagvikData

*Load Utqiagvik Temperature Data from GitHub Repository***Description**

This function retrieves the Utqiagvik Temperature dataset from a GitHub repository hosted at https://github.com/haghbinh/dataset/Rfssa_dataset. Hosting datasets on GitHub rather than including them in the Rfssa R package conserves storage space. The Utqiagvik Temperature dataset contains intraday hourly temperature curves measured in degrees Celsius from January 1973 to July 2023, recorded once per month. The returned object is a raw dataset in ‘list’ format;

Usage

```
loadUtqiagvikData()
```

Format

Containing two matrix objects:

Austin Temperature Data A 24 by 607 matrix of discrete samplings of intraday hourly temperature curves, one day per month.

Utqiagvik Temperature Data A 24 by 607 matrix of discrete samplings of intraday hourly temperature curves, one day per month.

References

Trinka, J., Haghbin, H., Shang, H., Maadooliat, M. (2023). Functional Time Series Forecasting: Functional Singular Spectrum Analysis Approaches. Stat, e621.

Examples

```
Utqiagvik_raw <- loadUtqiagvikData()
str(Utqiagvik_raw)
```

Montana

Montana Intraday Temperature Curves and NDVI Images Data Set

Description

This dataset includes intraday hourly temperature curves measured in degrees Celsius and normalized difference vegetation index (NDVI) image data. The observations were recorded near Saint Mary, Montana, USA. The NDVI images cover a geographical region with longitudes ranging from 113.30 degrees West to 113.56 degrees West and latitudes from 48.71 degrees North to 48.78 degrees North. The intraday temperature curves are sourced from Diamond et al. (2013), while the NDVI images were obtained from resources provided by Tuck et al. (2014). For each recorded intraday temperature curve, an NDVI image was captured on the same day every 16 days. Data collection started on January 1, 2008, and concluded on September 30, 2013. The primary goal of this dataset is to facilitate the analysis of temperature trends and investigate how temperature changes impact vegetation in the region. A multivariate analysis leveraging both temperature and NDVI variables can reveal more informative patterns and yield stronger signal extraction results. The dataset is hosted on GitHub, and you can load it using the function [loadMontanaData](#). The dataset has been converted into a multivariate functional time series using two B-spline basis function systems with 11 and 13 members, respectively. The resulting dataset is stored as a functional time series object of class ‘funts’. You can load the raw data using the function [loadMontanaData](#). See [funts](#) for more details.

References

1. Diamond, H. J., Karl, T., Palecki, M. A., Baker, C. B., Bell, J. E., Leeper, R. D., Easterling, D. R., Lawrimore, J. H., Meyers, T. P., Helfert, M. R., Goodge, G., and Thorne, P.W. (2013). U.S. climate reference network after one decade of operations: status and assessment. [Link](<https://www.ncdc.noaa.gov/crn/qcdatasets.html>). Last accessed April 2020.

2. Tuck, S. L., Phillips, H. R., Hintzen, R. E., Scharlemann, J. P., Purvis, A., and Hudson, L. N. (2014). MODISTools – downloading and processing MODIS remotely sensed data in R. Ecology and Evolution, 4(24), 4658–4668.

See Also

[loadMontanaData](#) - Function to load the Montana dataset.

plot.fforecast	<i>Plot Method for FSSA Forecast (fforecast) Class</i>
----------------	--

Description

Create visualizations of FSSA Forecast (fforecast) class. This function supports plotting ‘fforecast’ data with one-dimensional or two-dimensional domains.

Usage

```
## S3 method for class 'fforecast'
plot(
  x,
  group_index = NULL,
  ask = TRUE,
  npts = 100,
  obs = 1,
  main = NULL,
  col = NULL,
  ori_col = NULL,
  type = "l",
  lty = 1,
  ...
)
```

Arguments

x	an object of class fforecast.
group_index	an integer specifying the group index for the plot.
ask	logical: If ‘TRUE’, and ‘group_index’ be ‘NULL’, after printing the first grouping graphic, it will pause when the user asks for the next group graphic and wait.
npts	number of grid points for the plots.
obs	observation number (for two-dimensional domains).
main	main title for the plot.
col	specify the predicted FTS color; if it is ‘NULL’, it will be set as the default.
ori_col	specify the original FTS color; if it is ‘NULL’, it will be set as the default.

type	type of plot ("l" for line, "p" for points, etc.).
lty	line type (1 for solid, 2 for dashed, etc.).
...	additional graphical parameters passed to plotting functions.

See Also[fforecast](#)**Examples**

```
# Example with one-dimensional domain
data("Callcenter")
# FSSA Decomposition step:
fssa_results <- fssa(Callcenter, L = 28)

# Perform FSSA R-forecasting
pr_V <- fforecast(U = fssa_results, groups = list(1,1:7),
                  len = 14, method = "vector", only.new = FALSE)

plot(pr_V)
```

plot.fssa

*Plot Functional Singular Spectrum Analysis Objects***Description**

This function is a plotting method for objects of class functional singular spectrum analysis ([fssa](#)). It aids users in making decisions during the grouping stage of univariate or multivariate functional singular spectrum analysis.

Usage

```
## S3 method for class 'fssa'
plot(
  x,
  d = length(x$values),
  idx = 1:d,
  idy = idx + 1,
  contrib = TRUE,
  groups = as.list(1:d),
  lwd = 2,
  type = "values",
  vars = NULL,
  ylab = NA,
  main = NA,
  ...
)
```

Arguments

x	An object of class <code>fssa</code> .
d	An integer representing the number of elementary components to plot.
idx	A vector of indices specifying which eigen elements to plot.
idy	A second vector of indices of eigen elements to plot (for type="paired").
contrib	A logical value. If TRUE (default), it displays the component's contribution to the total variance.
groups	A list or vector of indices determining the grouping used for decomposition (for type="wcor").
lwd	A vector of line widths.
type	The type of plot to be displayed. Possible types include: • "values" - Plot the square-root of eigen values (default). • "paired" - Plot pairs of right singular function's coefficients (useful for detecting periodic components). • "wcor" - Plot the W-correlation matrix for the reconstructed objects. • "vectors" - Plot the right singular vectors (useful for detecting period length). • "lcurves" - Plot left singular functions (useful for detecting period length). • "lheats" - Heatmap plot of eigenfunctions, usable for <code>funts</code> variables observed over one or two-dimensional domains (useful for detecting meaningful patterns). • "periodogram" - Periodogram plot right singular vectors (useful for detecting the frequencies of oscillations in functional data).
vars	A numeric value specifying the variable number (used in plotting MFSSA "lheats" or "lcurves").
ylab	A character vector representing the names of variables.
main	The main plot title.
...	Additional arguments to be passed to methods, such as graphical parameters.

See Also

`fssa`, `plotly_funts`

Examples

```
data("Callcenter")
L <- 28
U <- fssa(Callcenter, L)
plot(U, type = "values", d = 10)
plot(U, type = "vectors", d = 4)
plot(U, type = "paired", d = 6)
plot(U, type = "lcurves", d = 4, vars = 1)
plot(U, type = "lheats", d = 4)
plot(U, type = "wcor", d = 10)
```

plot.funts	<i>Plot Functional Time Series (funts) Data</i>
------------	---

Description

Create visualizations of Functional Time Series (funts) data, supporting both one-dimensional and two-dimensional domains.

Usage

```
## S3 method for class 'funts'
plot(
  x,
  npts = 100,
  obs = 1,
  xlab = NULL,
  ylab = NULL,
  main = NULL,
  type = "l",
  lty = 1,
  ...
)
```

Arguments

x	An object of class funts.
npts	Number of grid points for the plots.
obs	Observation number (for two-dimensional domains).
xlab	X-axis label.
ylab	Y-axis label.
main	Main title for the plot.
type	Type of plot ("l" for line, "p" for points, etc.).
lty	Line type (1 for solid, 2 for dashed, etc.).
...	Additional graphical parameters passed to plotting functions.

Details

This function enables the creation of visualizations for Functional Time Series (funts) data. It supports both one-dimensional and two-dimensional domains.

For one-dimensional domains, line plots are used, while for two-dimensional domains, image plots are employed.

See Also

[funts](#), [Callcenter](#), [Montana](#)

Examples

```
# Example with one-dimensional domain
data("Callcenter")
plot(Callcenter, lwd = 2, col = "deepskyblue4", main = "Call Center Data")

# Example with two-dimensional domain
data("Montana")
plot(Montana, obs = 2, main = c("Temperature Curves", "NDVI Images,"))
```

plotly_funts

Plot Functional Time Series (funts) with Plotly

Description

Visualize univariate or multivariate Functional Time Series (funts) using Plotly-based plots.

Usage

```
plotly_funts(
  x,
  vars = NULL,
  types = NULL,
  subplot = TRUE,
  main = NULL,
  ylab = NULL,
  xlab = NULL,
  tlab = NULL,
  zlab = NULL,
  xticklabels = NULL,
  xticklocs = NULL,
  yticklabels = NULL,
  yticklocs = NULL,
  color_palette = "RdYlBu",
  reverse_color_palette = FALSE,
  ...
)
```

Arguments

x	an object of class funts .
vars	numeric vector specifying which variables in the FTS to plot (default: all).
types	tuple of strings specifying plot types for each variable.
subplot	logical for subplotting line plots.
main	titles for each plot.
ylab	y-axis titles.

xlab	x-axis titles.
tlab	time-axis titles.
zlab	z-axis titles.
xticklabels	tick labels for the domain of the functions.
xticklocs	positions of tick labels for the domain of the functions.
yticklabels	tick labels for the domain of the functions.
yticklocs	positions of tick labels for the domain of the functions.
color_palette	color palette for two-dimensional FTS plots.
reverse_color_palette	reverse the color palette scale.
...	additional arguments to pass to Plotly methods.

Details

Supported plot types for one-dimensional domain variables: - "line": line plots (default). - "heatmap": heatmaps. - "3Dsurface": 3D surface plots. - "3Dline": 3D line plots.

Supported plot type for two-dimensional domain variables: - "heatmap"

Each variable can be plotted multiple times with different types.

See Also

[funts](#), [Callcenter](#), [Montana](#)

Examples

```
## Not run:
data("Callcenter") # Univariate FTS example

plotly_funts(Callcenter)

plotly_funts(Callcenter,
  main = "Call Center Data Line Plot",
  xticklabels = list(c("00:00", "06:00", "12:00", "18:00", "24:00")),
  xticklocs = list(c(1, 60, 120, 180, 240))
)

plotly_funts(Callcenter, type = "3Dline", main = "Callcenter Data")

plotly_funts(Callcenter, type = "3Dsurface", main = "Callcenter Data")

plotly_funts(Callcenter, type = "heatmap", main = "Callcenter Data")

data("Montana") # Multivariate FTS example
plotly_funts(Montana[1:100],
  main = c("Temperature Curves", "NDVI Images"),
  color_palette = "RdYlGn",
  xticklabels = list(
```

```

      c("00:00", "06:00", "12:00", "18:00", "24:00"),
      c("113.40\u00B0 W", "113.30\u00B0 W")
    ),
    xticklocs = list(c(1, 6, 12, 18, 24), c(1, 33)),
    yticklabels = list(NA, c("48.70\u00B0 N", "48.77\u00B0 N")),
    yticklocs = list(NA, c(1, 33))
  )

## End(Not run)

```

print.fforecast	<i>Custom Print Method for FSSA Forecast (fforecast) class</i>
-----------------	--

Description

This custom print method is designed for objects of the FSSA Forecast (fforecast) class. It provides a summary of the fforecast object.

Usage

```

## S3 method for class 'fforecast'
print(x, ...)

```

Arguments

x	an object of class "fforecast" to be printed.
...	further arguments passed to or from other methods.

Examples

```

# Example with one-dimensional domain
data("Callcenter")
# FSSA Decomposition step:
fssa_results <- fssa(Callcenter, L = 28)

# Perform FSSA R-forecasting
pr_R <- fforecast(U = fssa_results,
                  groups = c(1:3),
                  len = 14,
                  method = "recurrent")

print(pr_R)

```

print.funts

Custom Print Method for Functional Time Series (funts) Objects

Description

This custom print method is designed for objects of the Functional Time Series (funts) class. It provides a summary of the funts object, including its length (N), the number of variables, and its structure.

Usage

```
## S3 method for class 'funts'
print(x, ...)
```

Arguments

x an object of class "funts" to be printed.
 ... further arguments passed to or from other methods.

Examples

```
## Not run:
data("Callcenter")
print(Callcenter)

## End(Not run)
```

[.funts

Indexing into Functional Time Series

Description

An indexing method for functional time series ([funts](#)) objects.

Usage

```
## S3 method for class 'funts'
obj[i = NULL, j = NULL]
```

Arguments

obj an object of class [funts](#).
 i an index or indices specifying the subsets of times to extract.
 j an index or indices specifying the subsets of variables to extract.

Details

This function allows you to extract specific subsets of a functional time series based on the provided indices. You can specify which subsets you want to extract from the functional time series.

Value

an 'funts' object containing the specified subsets

See Also

[funts](#)

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