Package 'GFPCA'

December 26, 2023

Title Graphical Functional Principal Component Analysis

Version 1.0

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Description With the development of data collection methods, multivariate functional data that possess complex temporal correlation structures have become increasingly available. To analyze these data, we need to account for both multivariate and temporal correlations. These two-way dependencies may lead to inefficient dimension reduction when using classical FPCA methods. To solve this issue, we propose a novel graphical FPCA (GFPCA) for the dimension reduction of multivariate functional time series (MFTS) data, in which the multivariate dependencies are characterized by graphical models. In this package, the GFPCA() function is used for optimal reconstructions of signals from contaminated MFTS data using graphical-level information. In addition, the function can also be used for capturing and encoding the graphical structures of MFTS.

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Encoding UTF-8

RoxygenNote 7.2.3

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eigen_func_dyn	Estimating Functional Filters
eigen_runc_uyn	Estimating Functional Fillers

Description

A function to estimate functional filters for dynamic FPCA.

Usage

```
eigen_func_dyn(sp_matrix, grid_freq, sel_eig, max_comp)
```

Arguments

sp_matrix	An array of the estimated coefficients for spectral density kernels.
grid_freq	The frequencies of the estimated spectral density kernels.
sel_eig	The fraction of variance explained.
max_comp	The maximum number of components.

Value

Estimations of the number of components and the coefficients of functional filters.

eigen_func_sta	Estimating Eigenfunctions	

Description

A function to estimate eigenfunctions for static FPCA.

Usage

```
\verb|eigen_func_sta|(\verb|cov_matrix_0|, \verb|sel_eig|, \verb|max_comp|)|
```

Arguments

cov_matrix_0	The estimated coefficients for lag-zero covariance functions.
sel_eig	The fraction of variance explained.
max_comp	The maximum number of components.

Value

Estimations of the number of components and the coefficients of eigenfunctions.

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eigen_matrix_con

Estimating Eigen-matrices

Description

A function to estimate eigen-matrices.

Usage

```
eigen_matrix_con(
  time_length,
  comp_num,
  subject_length,
  dmean_smo_fda,
  q,
  r,
  cov_matrix,
  Freq_eig,
  time_length_eig
)
```

Arguments

```
time_length The time length of MFTS.

comp_num The number of components.

subject_length The dimension of MFTS.

dmean_smo_fda The pre-smoothed coefficients after removing mean trends.

cov_matrix An array of the estimated coefficients for the cross-covariance functions.

Freq_eig The frequencies for the estimated eigenmatrices.

time_length_eig The number of the evaluated frequencies.

p, r The time lags for the estimated covariance functions.
```

Value

An array of the estimated eigenmatrices.

eigen_pre

Estimating Cross-covariance

Description

A function to estimate cross-covariance for different time lags.

Usage

```
eigen_pre(subject_length, time_length, dmean_smo_fda, q, r)
```

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Arguments

subject_length The dimension of MFTS.

time_length The time length of MFTS.

dmean_smo_fda The pre-smoothed coefficients after removing mean trends.

p, r The time lags for the estimated covariance functions.

Value

An array of the estimated coefficients for the cross-covariance functions.

GFPCA Graphical Functional Principal Component Analysis (GFPCA)

Description

A function to implement graphical FPCAs.

Usage

```
GFPCA(MFTS, Lt, Dynamic, Max.comp, FVE, Mean.zero)
```

Arguments

MFTS	A Z*p*J array containing the regularly observed MFTS data, where Z is the number of observed time points, p is the dimension of MFTS, and J is the time length of MFTS.
Lt	A vector containing the observed time points of MFTS. We require the time points to be contained in the interval [0,1].
Dynamic	A logical evaluating to TRUE or FALSE indicating whether the graphical DF-PCA or SFPCA should be conducted.
Max.comp	The maximum number of components.
FVE	A numeric in $[0,1]$ indicating the fraction of variance explained for determining the number of components. If $FVE = 0$, the number of component is selected by the ratio of variance explained.
Mean.zero	A logical evaluating to TRUE or FALSE indicating whether the mean functions of MFTS are zero.

Details

This is a generic function to implement graphical versions of DFPCA or SFPCA for multivariate functional time series (MFTS). The function requires that the MFTS data are densely and regularly observed.

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Value

A list with components

comp_num Number of components.

xi_dyn_IN or xi_sta_IN

A list of FPC scores computed by integration.

xi_dyn_CE or xi_sta_CE

A list of FPC scores computed by conditional expectation.

mean_function A list of mean functions of MFTS, where Lt contains the time grid of the esti-

mated values.

Functional filters or eigenfunctions

A list of the estimated functional filters (or eigenfunctions), where Lt contains

the time grid of the estimated values.

eigen_matrix A p*p*(J/2)*comp_num array containing the estimated eigenmatrices for dif-

ferent frequences.

Phi A p*p*(J/2)*comp_num array containing the estimated inverse of eigenmatrices

by incorporating graph constraints.

mea_error A vector containing the estimated variances of the measurement errors.

gra_selection Regularized Estimator for Precision Matrices

Description

Joint regularized estimator for precision matrices by ADMM.

Usage

gra_selection(S, lambda)

Arguments

S An array of eigenmatrices.

lambda A tuning parameter to the joint regularized estimator.

Value

A list of two collections of precision matrices.

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score_est

Score Extraction

Description

A function to extract FPCA scores for Graphical FPCA.

Usage

```
score_est(
   xi,
   fre_tra,
   des_mat,
   dmean_fda,
   eigen_matrix,
   squ_des_mat,
   squ_fre_tra,
   sigma_est,
   subject_length,
   time_length,
   comp_num
)
```

Arguments

Value

A list containing the estimated scores and other objects relating to the convergence of the algorithm.

Sel_lam

Selection of Tuning Parameters for Joint Graphical Lasso

Description

A function to select tuning parameters for joint graphical Lasso via AIC.

Usage

```
Sel_lam(eigen_matrix)
```

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Arguments

eigen_matrix An array of eigenmatrices.

Value

A selected tuning parameter.

smooth_fda

Pre-smoothing of Multivariate Functional Time Series (MFTS)

Description

A function to pre-smooth MFTS data.

Usage

```
smooth_fda(fda, x_fda, subject_length, time_length, X_basis, tra_mat)
```

Arguments

fda An array containing the observed MFTS data.

x_fda The observed time points of MFTS.

 $\begin{tabular}{ll} subject_length & The dimension of MFTS. \\ time_length & The time length of MFTS. \\ \end{tabular}$

X_basis A basis object for pre-smoothing.

tra_mat A transformation matrix for the basis functions.

Value

A list with the pre-smoothed coefficients and the estimated variances of measurement errors.

sp_tran Constructing Spectral Density Kernels

Description

A function to construct spectral densities given cross-covariance.

Usage

```
sp_tran(r, grid_freq, cov_matrix)
```

Arguments

The time lags for estimating the spectral density kernels.

grid_freq The frequencies of the estimated spectral density kernels.

cov_matrix An array of the estimated coefficients for the cross-covariance functions.

Value

An array of the estimated coefficients for spectral density kernels.

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test_weaksep

Testing Dynamic Weak Separability

Description

A function to illustrate the validity of dynamic weak separability.

Usage

```
test_weaksep(
   time_length,
   comp_num,
   subject_length,
   dmean_smo_fda,
   q,
   r,
   Freq_eig,
   time_length_eig
)
```

Arguments

```
time_length The time length of MFTS.

comp_num The number of components.

subject_length The dimension of MFTS.

dmean_smo_fda The pre-smoothed coefficients after removing mean trends.

Freq_eig The frequencies for the estimated eigenmatrices.

time_length_eig The number of the evaluated frequencies.

p, r The time lags for the estimated covariance functions.
```

Value

An object to illustrate the validity of dynamic weak separability.

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