Package 'ctmva'

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Type Package

Title Continuous-Time Multivariate Analysis

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Description Implements a basis function or functional data analysis framework
for several techniques of multivariate analysis in continuous-time
setting. Specifically, we introduced continuous-time analogues of
several classical techniques of multivariate analysis, such as
principal component analysis, canonical correlation analysis,
Fisher linear discriminant analysis, K-means clustering, and so
on. Details are in Philip T Reiss and Biplab Paul (2022) ``Continuous-time multivariate analysis";
James O Ramsay, Bernard W Silverman (2005) <ISBN:978-0-387-22751-
1> ``Functional Data Analysis";
James O Ramsay, Giles Hooker and Spencer Graves (2009) <ISBN:978-0-387-98185-
7> ``Functional Data Analysis with R and MATLAB".

License GPL (>= 2)

Imports fda, polynom

Suggests eegkit, corrplot

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`ctmva-package` *Continuous-time multivariate analysis*

Description

Implements continuous-time analogues of several classical techniques of multivariate analysis. The inputs are `"fd"` (functional data) objects from the **fda** package.

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

References

Reiss, Philip T. and Paul, Biplab (2022). Continuous-time multivariate analysis. Preprint.

Ramsay, James O. and Silverman, Bernard W. (2005). Functional Data Analysis, 2nd ed., Springer, New York.

Ramsay, J. O.; Hooker, Giles and Graves, Spencer (2009). Functional Data Analysis with R and Matlab, Springer, New York.

`cca.ct` *Continuous-time canonical correlation analysis*

Description

A continuous-time version of canonical correlation analysis (CCA).

Usage

```
cca.ct(fdobj1, fdobj2)
```

Arguments

`fdobj1, fdobj2`	a pair of continuous-time multivariate data sets, of class `"fd"`

Value

A list consisting of

vex1, vex2	matrices defining the canonical variates. The first columns of each give the coefficients defining the first pair of canonical variates; and so on.
cor	canonical correlations, i.e., correlations between the pairs of canonical variates

Note

Columns of the output matrix vex2 are flipped as needed to ensure positive correlations.

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

cancor, for classical CCA

Examples

```
# CCA relating Canadian daily temperature and precipitation data
require(fda)
data(CanadianWeather)
daybasis <- create.bspline.basis(c(0,365), nbasis=80)
tempfd <- smooth.basis(day.5, CanadianWeather$dailyAv[,,"Temperature.C"], daybasis)$fd
precfd <- smooth.basis(day.5, CanadianWeather$dailyAv[,,"log10precip"], daybasis)$fd
tpcor <- cca.ct(tempfd, precfd)
oldpar <- par(mfrow=1:2)
barplot(tpcor$vex1[,1], horiz=TRUE, las=1, main="Temperature",
            sub="First canonical coefficients vector")
barplot(tpcor$vex2[,1], horiz=TRUE, las=1, main="Log precipitation",
            sub="First canonical coefficients vector")
par(oldpar)
```

---

cor.ct *Continuous-time correlation or cross-correlation matrix*

Description

Computes the correlation matrix of a continuous-time multivariate data set represented as an fd object; or the cross-correlation matrix of two such data sets.

Usage

```
cor.ct(fdobj1, fdobj2 = fdobj1, common_trend = FALSE)
```

Arguments

fdobj1	continuous-time multivariate data set of class "fd"
fdobj2	an optional second data set
common_trend	logical: centering wrt mean function if TRUE, without centering if FALSE (the default)

Value

A matrix of (cross-) correlations

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

center.fd, for centering of "fd" objects; inprod.cent

Examples

```
# Canadian temperature data

require(fda)
require(corrplot)
data(CanadianWeather)
daybasis <- create.fourier.basis(c(0,365), nbasis=55)
tempfd <- smooth.basis(day.5, CanadianWeather$dailyAv[,,"Temperature.C"], daybasis)$fd

## The following yields a matrix of correlations that are all near 1:
rawcor <- cor.ct(tempfd)
corrplot(rawcor, method = 'square', type = 'lower', tl.col="black", tl.cex = 0.6)
## This occurs due to a strong seasonal trend that is common to all stations
## Removing this common trend leads to a more interesting result:
dtcor  <- cor.ct(tempfd, common_trend = TRUE)
ord <- corrMatOrder(dtcor)
dtcord <- dtcor[ord,ord]
corrplot(dtcord, method = 'square', type = 'lower', tl.col="black", tl.cex = 0.6)
```

---

cov.ct *Continuous-time covariance or cross-covariance matrix*

Description

Computes the covariance matrix of a continuous-time multivariate data set represented as an fd object; or the cross-covariance matrix of two such data sets.

Usage

```
cov.ct(fdobj1, fdobj2 = fdobj1, common_trend = FALSE)
```

Arguments

fdobj1	continuous-time multivariate data set of class "fd"
fdobj2	an optional second data set
common_trend	logical: centering with respect to the mean function if TRUE, without centering if FALSE (the default)

Value

A matrix of (cross-) covariances

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

cor.ct

Examples

```
# see example for cor.ct, which works similarly
```

---

inprod.cent *Centered inner product matrix for a basis or pair of bases*

Description

Most methods of continous-time multivariate analysis require a matrix of inner products of pairs of functions from a basis, such as a B-spline basis, or pairs consisting of one function from each of two bases. This function computes such matrices via 7-point Newton-Cotes integration, which is exact for cubic B-splines.

Usage

```
inprod.cent(basis1, basis2 = basis1, rng = NULL)
```

Arguments

basis1	basis object from the fda package.
basis2	an optional second basis
rng	range (of times) spanned by the basis

Value

Matrix of inner products of each pair of centered basis functions.

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

`create.bspline.basis` from package `fda`, for the most commonly used basis object type.

Examples

```
require(fda)
basis8 <- create.bspline.basis(nbasis=8)
inprod.cent(basis8)
```

---

kmeans.ct *Continuous-time k-means clustering*

Description

A continuous-time version of k-means clustering in which each clusters is a time segments or set of time segments.

Usage

```
kmeans.ct(
  fdobj,
  k,
  common_trend = FALSE,
  init.pts = NULL,
  tol = 0.001,
  max.iter = 100
)
```

Arguments

`fdobj`	continuous-time multivariate data set of class "`fd`"
`k`	number of clusters
`common_trend`	logical: Should the curves be centered with respect to the mean function? Defaults to `FALSE`.
`init.pts`	a set of k time points. The observations at these time points serve as initial values for the k means. Randomly generated if not supplied.

tol	convergence tolerance for the k means
max.iter	maximum number of iterations

Value

Object of class "kmeans.ct", a list consisting of

fdobj	the supplied fdobj
means	means of the k clusters
transitions	transition points between segments
cluster	cluster memberships in the segments defined by the transitions

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

plot.kmeans.ct

Examples

```
require(fda)
data(CanadianWeather)
daybasis <- create.bspline.basis(c(0,365), nbasis=55)
tempfd <- smooth.basis(day.5, CanadianWeather$dailyAv[,,"Temperature.C"], daybasis)$fd
kmtemp3 <- kmeans.ct(tempfd, 3)
plot(kmtemp3)
```

---

lda.ct *Continuous-time Fisher's linear discriminant analysis*

Description

A continuous-time version of Fisher's LDA, in which segments of the time interval take the place of groups of observations.

Usage

```
## S3 method for class 'ct'
lda(fdobj, partition)
```

Arguments

fdobj	continuous-time multivariate data set of class "fd"
partition	a priori break points dividing the time interval into segments

Value

Object of class "lda.ct", a list consisting of

scaling	matrix of coefficients defining the discriminants (as in lda)
values	eigenvalues giving the ratios of between to within sums of squares
partition	the supplied partition
fdobj	linear discriminants represented as an "fd" object
nld	number of linear discriminants

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

plot.lda.ct; lda, for the classical version

Examples

```
## see end of example in ?pca.ct
```

---

meanbasis *Compute means of basis functions*

Description

Given a basis object as defined in the **fda** package (see basisfd), this function simply computes the vector of means of the basis functions. Used internally.

Usage

```
meanbasis(basis)
```

Arguments

basis	a basis object of class "basisfd"

Value

Vector of means of the basis functions

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

Examples

```
require(fda)
fbasis11 <- create.fourier.basis(nbasis=11)
zapsmall(meanbasis(fbasis11))    # the sine functions have mean 0
```

---

pca.ct *Continuous-time principal component analysis*

Description

A continuous-time version of principal component analysis.

Usage

```
pca.ct(fdobj, cor = FALSE, common_trend = FALSE)
```

Arguments

fdobj	continuous-time multivariate data set of class "fd"
cor	logical: use correlation matrix if TRUE, covariance if FALSE (the default)
common_trend	logical: Should the curves be centered with respect to the mean function? Defaults to FALSE.

Value

Returns a list including:

var	variances of the principal components.
loadings	the matrix of loadings (i.e., its columns are the eigenvectors of the continuous-time covariance).
scorefd	score functions.

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

cov.ct; princomp, for the classical version

Examples

```
# Data for one session from a classic EEG data set
require(fda)
require(eegkit)
data(eegdata)
data(eegcoord)
longdat <- subset(eegdata, subject=="co2a0000369" & trial==0)
widedat <- reshape(longdat, direction="wide", drop=c("subject","group","condition","trial"),
                v.names="voltage",idvar="channel")

# Convert time series for 64 channels to a functional data object
bsb <- create.bspline.basis(c(0,255),nbasis=30)
fdo <- Data2fd(argvals=0:255, y=t(as.matrix(widedat[,-1])), basisobj=bsb)
plot(fdo)

# Now do PCA and display first loadings for 3 PC's,
# along with percent variance explained by each
pcc <- pca.ct(fdo)
pve <- 100*pcc$var/sum(pcc$var)
oldpar <- par(mfrow=c(1,3))
cidx <- match(widedat[,1],rownames(eegcoord))
eegspace(eegcoord[cidx,4:5],pcc$loadings[,1], colorlab="PC1 loadings",
         main=paste0(round(pve[1],0), "%"), mar=c(17,3,12,2), cex.main=2)
eegspace(eegcoord[cidx,4:5],pcc$loadings[,2], colorlab="PC2 loadings",
         main=paste0(round(pve[2],0), "%"), mar=c(17,3,12,2), cex.main=2)
eegspace(eegcoord[cidx,4:5],pcc$loadings[,3], colorlab="PC3 loadings",
         main=paste0(round(pve[3],0), "%"), mar=c(17,3,12,2), cex.main=2)

# Linear discriminant analysis: discriminating among the 1st, 2nd and 3rd portions
#  of the time interval
ld <- lda.ct(fdo, c(85,170))
plot(ld)
eegspace(eegcoord[cidx,4:5],ld$scaling[,1], colorlab="LD1 coefficients",
         mar=c(17,3,12,2), cex.main=2)
eegspace(eegcoord[cidx,4:5],ld$scaling[,2], colorlab="LD2 coefficients",
         mar=c(17,3,12,2), cex.main=2)
par(oldpar)
```

`plot.kmeans.ct` *Plot a kmeans.ct object*

Description

Plots a continuous-time k-means clustering object generated by a call to `kmeans.ct`.

Usage

```
## S3 method for class 'kmeans.ct'
plot(
  x,
  type = "functions",
  mark.transitions = TRUE,
  col = NULL,
  lty = NULL,
  xlab = "Time",
  ylab = NULL,
  legend = TRUE,
  ncol.legend = 1,
  ...
)
```

Arguments

x	clustering object produced by kmeans.ct
type	either "functions" (the default), to display each variable as a smooth function of time, or "distance", to plot distances from the k cluster means versus time.
mark.transitions	logical: Should transitions between clusters be marked with vertical lines? Defaults to TRUE.
col	plot colours
lty	line type
xlab, ylab	x- and y-axis labels
legend	logical: should a legend be included? Default is TRUE.
ncol.legend	number of columns for legend
...	other arguments passed to matplot

Value

None; a plot is generated.

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

kmeans.ct, which includes an example

`plot.lda.ct` *Plot an lda.ct object*

Description

Plots the Fisher's linear discriminant functions generated by a call to `lda.ct`.

Usage

```
## S3 method for class 'lda.ct'
plot(x, ylab = "Discriminants", xlab = "Time", ...)
```

Arguments

`x`	linear discriminant analysis object produced by `lda.ct`
`ylab, xlab`	y- and x-axis labels
`...`	other arguments passed to `matplot`

Value

None; a plot is generated.

Author(s)

Biplab Paul <paul.biplab497@gmail.com> and Philip Tzvi Reiss <reiss@stat.haifa.ac.il>

See Also

`lda.ct`

Examples

```
## see the example at the end of ?pca.ct
```

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