

Package ‘logitFD’

May 8, 2026

Type Package

Title Functional Principal Components Logistic Regression

Version 1.0

Date 2021-12-22

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Description Functions for fitting a functional principal components logit regression model in four different situations: ordinary and filtered functional principal components of functional predictors, included in the model according to their variability explanation power, and according to their prediction ability by stepwise methods. The proposed methods were developed in Escabias et al (2004) <[doi:10.1080/10485250310001624738](https://doi.org/10.1080/10485250310001624738)> and Escabias et al (2005) <[doi:10.1016/j.csda.2005.03.011](https://doi.org/10.1016/j.csda.2005.03.011)>.

Depends R (>= 3.4.3)

License GPL (>= 2)

Imports fda, expm, pROC, fda.usc

Encoding UTF-8

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NeedsCompilation no

Repository CRAN

Date/Publication 2022-01-10 08:52:43 UTC

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logitFD.fpc	<i>Filtered Functional Principal Component Logistic Regression by explained variability order</i>
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Description

Fit of the Filtered Functional Principal Component Logistic Regression model with selected Functional Principal Components included in the model according their explained variability.

Usage

```
logitFD.fpc(Response, FDObj = list(), ncomp = c(), nonFDvars = NULL)
```

Arguments

Response	Binary (numeric or character) vector of observations of the response variable.
FDObj	List of functional objects from fda package with the curves of the predictor functional variables.
ncomp	Numeric vector with the number of components to be considered for each functional predictor. The vector has equal length than FDObj.
nonFDvars	Matrix or data frame with the observations of non-functional variables.

Value

glm.fit	glm object of the fitted model. The object allows to use the summary() function for printing a summary of the fit, the anova() function to produce an analysis of variance table, and to extract useful features as coefficients, effects, fitted.values or residuals
Intercept	Intercept estimated parameter
betalist	List of functional objects (fdoj) of fda package with the estimated parameter functions. Each element of the list corresponds to the associated functional predictor located in the same position of FDObj. All methods of fd package can be used as the plot() function among others.
PC.variance	List of data frames with explained variability of functional principal components of functional predictors. Each element of the list contains the acumulation variance matrix corresponding to the associated functional variable in the same position.
ROC.curve	Object of the roc() function of the pROC package for prediction ability testing of the model. The object can be printed, plotted, or passed to many other functions

Author(s)

Carmen Lucia Reina <carmenlureina@gmail.com> and Manuel Escabias <escabias@ugr.es>

Examples

```

library(fda.usc)
data(aemet)
Temp<-aemet$temp$data
Prec<-exp(aemet$logprec$data)
StationsVars<-aemet$df[,c("ind", "altitude", "longitude", "latitude")]
StationsVars$North<-c(1,1,1,1,0,0,0,0,1,1,0,0,0,0,1,1,1,0,0,1,0,0,0,0,1,0,0,1,1,1,1,1,
0,0,0,1,1,1,1,1,1,1,0,0,0,0,1,1,1,1,1,0,0,0,0,0,0,0,0,1,1,1,0,0,1,1,1,1,1,1)
TempMonth<-matrix(0,73,12)
PrecMonth<-matrix(0,73,12)
for (i in 1:nrow(TempMonth)){
  TempMonth[i,1]<-mean(Temp[i,1:31])
  PrecMonth[i,1]<-mean(Prec[i,1:31])
  TempMonth[i,2]<-mean(Temp[i,32:59])
  PrecMonth[i,2]<-mean(Prec[i,32:59])
  TempMonth[i,3]<-mean(Temp[i,60:90])
  PrecMonth[i,3]<-mean(Prec[i,60:90])
  TempMonth[i,4]<-mean(Temp[i,91:120])
  PrecMonth[i,4]<-mean(Prec[i,91:120])
  TempMonth[i,5]<-mean(Temp[i,121:151])
  PrecMonth[i,5]<-mean(Prec[i,121:151])
  TempMonth[i,6]<-mean(Temp[i,152:181])
  PrecMonth[i,6]<-mean(Prec[i,152:181])
  TempMonth[i,7]<-mean(Temp[i,182:212])
  PrecMonth[i,7]<-mean(Prec[i,182:212])
  TempMonth[i,8]<-mean(Temp[i,213:243])
  PrecMonth[i,8]<-mean(Prec[i,213:243])
  TempMonth[i,9]<-mean(Temp[i,244:273])
  PrecMonth[i,9]<-mean(Prec[i,244:273])
  TempMonth[i,10]<-mean(Temp[i,274:304])
  PrecMonth[i,10]<-mean(Prec[i,274:304])
  TempMonth[i,11]<-mean(Temp[i,305:334])
  PrecMonth[i,11]<-mean(Prec[i,305:334])
  TempMonth[i,12]<-mean(Temp[i,335:365])
  PrecMonth[i,12]<-mean(Prec[i,335:365])
}
FourierBasis<-create.fourier.basis(rangeval = c(1,12),nbasis=7)
BsplineBasis<-create.bspline.basis(rangeval = c(1,12),nbasis=8)
TempMonth.fd<-Data2fd(argvals = c(1:12), y=t(TempMonth),basisobj = FourierBasis)
PrecMonth.fd<-Data2fd(argvals = c(1:12), y=t(PrecMonth),basisobj = BsplineBasis)
logitFD.fpc(Response=StationsVars$North,FDobj=list(TempMonth.fd,PrecMonth.fd),
  ncomp = c(3,4),nonFDvars = StationsVars[,c("altitude","longitude")])

```

logitFD.fpc.step

*Filtered Functional Principal Component Logistic Regression by step-wise order***Description**

Fit of the Filtered Functional Principal Component Logistic Regression model with Functional Principal Components and nonfunctional variables included in the model according their prediction

ability by an automatic stepwise selection method.

Usage

```
logitFD.fpc.step(Response, FDObj = list(), nonFDvars = NULL)
```

Arguments

Response	Binary (numeric or character) vector of observations of the response variable.
FDObj	List of functional objects from fda package with the curves of the predictor functional variables.
nonFDvars	Matrix or data frame with the observations of non-functional variables.

Value

glm.fit	glm object of the fitted model. The object allows to use the summary() function for printing a summary of the fit, the anova() function to produce an analysis of variance table, and to extract useful features as coefficients, effects, fitted.values or residuals
Intercept	Intercept estimated parameter
betalist	List of functional objects (fdobj) of fda package with the estimated parameter functions. Each element of the list corresponds to the associated functional predictor located in the same position of FDObj. All methods of fd package can be used as the plot() function among others.
PC.variance	List of data frames with explained variability of functional principal components of functional predictors. Each element of the list contains the accumulation variance matrix corresponding to the associated functional variable in the same position.
ROC.curve	Object of the roc() function of the pROC package for prediction ability testing of the model. The object can be printed, plotted, or passed to many other functions

Author(s)

Carmen Lucia Reina <carmenlureina@gmail.com> and Manuel Escabias <escabias@ugr.es>

Examples

```
library(fda.usc)
data(aemet)
Temp<-aemet$temp$data
Prec<-exp(aemet$logprec$data)
StationsVars<-aemet$df[,c("ind", "altitude", "longitude", "latitude")]
StationsVars$North<-c(1,1,1,1,0,0,0,0,1,1,0,0,0,0,0,1,1,1,0,0,1,0,0,0,0,1,0,0,1,1,1,1,1,
0,0,0,1,1,1,1,1,1,1,1,0,0,0,0,1,1,1,1,1,0,0,0,0,0,0,0,0,1,1,1,0,0,1,1,1,1,1,1)
TempMonth<-matrix(0,73,12)
PrecMonth<-matrix(0,73,12)
for (i in 1:nrow(TempMonth)){
  TempMonth[i,1]<-mean(Temp[i,1:31])
  PrecMonth[i,1]<-mean(Prec[i,1:31])
}
```

```

TempMonth[i,2]<-mean(Temp[i,32:59])
PrecMonth[i,2]<-mean(Prec[i,32:59])
TempMonth[i,3]<-mean(Temp[i,60:90])
PrecMonth[i,3]<-mean(Prec[i,60:90])
TempMonth[i,4]<-mean(Temp[i,91:120])
PrecMonth[i,4]<-mean(Prec[i,91:120])
TempMonth[i,5]<-mean(Temp[i,121:151])
PrecMonth[i,5]<-mean(Prec[i,121:151])
TempMonth[i,6]<-mean(Temp[i,152:181])
PrecMonth[i,6]<-mean(Prec[i,152:181])
TempMonth[i,7]<-mean(Temp[i,182:212])
PrecMonth[i,7]<-mean(Prec[i,182:212])
TempMonth[i,8]<-mean(Temp[i,213:243])
PrecMonth[i,8]<-mean(Prec[i,213:243])
TempMonth[i,9]<-mean(Temp[i,244:273])
PrecMonth[i,9]<-mean(Prec[i,244:273])
TempMonth[i,10]<-mean(Temp[i,274:304])
PrecMonth[i,10]<-mean(Prec[i,274:304])
TempMonth[i,11]<-mean(Temp[i,305:334])
PrecMonth[i,11]<-mean(Prec[i,305:334])
TempMonth[i,12]<-mean(Temp[i,335:365])
PrecMonth[i,12]<-mean(Prec[i,335:365])
}
FourierBasis<-create.fourier.basis(rangeval = c(1,12),nbasis=7)
BsplineBasis<-create.bspline.basis(rangeval = c(1,12),nbasis=8)
TempMonth.fd<-Data2fd(argvals = c(1:12), y=t(TempMonth),basisobj = FourierBasis)
PrecMonth.fd<-Data2fd(argvals = c(1:12), y=t(PrecMonth),basisobj = BsplineBasis)
logitFD.fpc.step(Response=StationsVars$North,FDobj=list(TempMonth.fd,PrecMonth.fd),
nonFDvars = StationsVars[,c("altitude","longitude")])

```

logitFD.pc

*Functional Principal Component Logistic Regression by explained
variability order*

Description

Fit of the Functional Principal Component Logistic Regression model with selected Functional Principal Components included in the model according their explained variability.

Usage

```
logitFD.pc(Response, FDobj = list(), ncomp = c(), nonFDvars = NULL)
```

Arguments

Response	Binary (numeric or character) vector of observations of the response variable.
FDobj	List of functional objects from fda package with the curves of the predictor functional variables.

ncomp	Numeric vector with the number of components to be considered for each functional predictor. The vector has equal length than FDObj.
nonFDvars	Matrix or data frame with the observations of non-functional variables.

Value

glm.fit	glm object of the fitted model. The object allows to use the summary() function for printing a summary of the fit, the anova() function to produce an analysis of variance table, and to extract useful features as coefficients, effects, fitted.values or residuals
Intercept	Intercept estimated parameter
betalist	List of functional objects (fdoj) of fda package with the estimated parameter functions. Each element of the list corresponds to the associated functional predictor located in the same position of FDObj. All methods of fd package can be used as the plot() function among others.
PC.variance	List of data frames with explained variability of functional principal components of functional predictors. Each element of the list contains the accumulation variance matrix corresponding to the associated functional variable in the same position.
ROC.curve	Object of the roc() function of the pROC package for prediction ability testing of the model. The object can be printed, plotted, or passed to many other functions

Author(s)

Carmen Lucia Reina <carmenlureina@gmail.com> and Manuel Escabias <escabias@ugr.es>

Examples

```
library(fda.usc)
data(aemet)
Temp<-aemet$temp$data
Prec<-exp(aemet$logprec$data)
StationsVars<-aemet$df[,c("ind", "altitude", "longitude", "latitude")]
StationsVars$North<-c(1,1,1,1,0,0,0,0,1,1,0,0,0,0,1,1,1,0,0,1,0,0,0,0,1,0,0,1,1,1,1,1,
0,0,0,1,1,1,1,1,1,1,0,0,0,0,1,1,1,1,1,0,0,0,0,0,0,0,1,1,1,0,0,1,1,1,1,1,1)
TempMonth<-matrix(0,73,12)
PrecMonth<-matrix(0,73,12)
for (i in 1:nrow(TempMonth)){
  TempMonth[i,1]<-mean(Temp[i,1:31])
  PrecMonth[i,1]<-mean(Prec[i,1:31])
  TempMonth[i,2]<-mean(Temp[i,32:59])
  PrecMonth[i,2]<-mean(Prec[i,32:59])
  TempMonth[i,3]<-mean(Temp[i,60:90])
  PrecMonth[i,3]<-mean(Prec[i,60:90])
  TempMonth[i,4]<-mean(Temp[i,91:120])
  PrecMonth[i,4]<-mean(Prec[i,91:120])
  TempMonth[i,5]<-mean(Temp[i,121:151])
  PrecMonth[i,5]<-mean(Prec[i,121:151])
  TempMonth[i,6]<-mean(Temp[i,152:181])
}
```

```

PrecMonth[i,6]<-mean(Prec[i,152:181])
TempMonth[i,7]<-mean(Temp[i,182:212])
PrecMonth[i,7]<-mean(Prec[i,182:212])
TempMonth[i,8]<-mean(Temp[i,213:243])
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TempMonth[i,11]<-mean(Temp[i,305:334])
PrecMonth[i,11]<-mean(Prec[i,305:334])
TempMonth[i,12]<-mean(Temp[i,335:365])
PrecMonth[i,12]<-mean(Prec[i,335:365])
}
FourierBasis<-create.fourier.basis(rangeval = c(1,12),nbasis=7)
BsplineBasis<-create.bspline.basis(rangeval = c(1,12),nbasis=8)
TempMonth.fd<-Data2fd(argvals = c(1:12), y=t(TempMonth),basisobj = FourierBasis)
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logitFD.pc(Response=StationsVars$North,FDobj=list(TempMonth.fd,PrecMonth.fd),
  ncomp = c(3,4),nonFDvars = StationsVars[,c("altitude","longitude")])

```

logitFD.pc.step	<i>Functional Principal Component Logistic Regression by stepwise order</i>
-----------------	---

Description

Fit of the Functional Principal Component Logistic Regression model with Functional Principal Components and nonfunctional variables included in the model according their prediction ability by an automatic stepwise selection method.

Usage

```
logitFD.pc.step(Response, FDobj = list(), nonFDvars = NULL)
```

Arguments

Response	Binary (numeric or character) vector of observations of the response variable.
FDobj	List of functional objects from fda package with the curves of the predictor functional variables.
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Intercept	Intercept estimated parameter
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Examples

```
library(fda.usc)
data(aemet)
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StationsVars<-aemet$df[,c("ind", "altitude", "longitude", "latitude")]
StationsVars$North<-c(1,1,1,0,0,0,0,1,1,0,0,0,0,0,1,1,1,0,0,1,0,0,0,0,1,0,0,1,1,1,1,1,
0,0,0,1,1,1,1,1,1,1,0,0,0,0,1,1,1,1,1,0,0,0,0,0,0,0,0,1,1,1,0,0,1,1,1,1,1)
TempMonth<-matrix(0,73,12)
PrecMonth<-matrix(0,73,12)
for (i in 1:nrow(TempMonth)){
  TempMonth[i,1]<-mean(Temp[i,1:31])
  PrecMonth[i,1]<-mean(Prec[i,1:31])
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  PrecMonth[i,2]<-mean(Prec[i,32:59])
  TempMonth[i,3]<-mean(Temp[i,60:90])
  PrecMonth[i,3]<-mean(Prec[i,60:90])
  TempMonth[i,4]<-mean(Temp[i,91:120])
  PrecMonth[i,4]<-mean(Prec[i,91:120])
  TempMonth[i,5]<-mean(Temp[i,121:151])
  PrecMonth[i,5]<-mean(Prec[i,121:151])
  TempMonth[i,6]<-mean(Temp[i,152:181])
  PrecMonth[i,6]<-mean(Prec[i,152:181])
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  TempMonth[i,8]<-mean(Temp[i,213:243])
  PrecMonth[i,8]<-mean(Prec[i,213:243])
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  PrecMonth[i,9]<-mean(Prec[i,244:273])
  TempMonth[i,10]<-mean(Temp[i,274:304])
  PrecMonth[i,10]<-mean(Prec[i,274:304])
  TempMonth[i,11]<-mean(Temp[i,305:334])
  PrecMonth[i,11]<-mean(Prec[i,305:334])
  TempMonth[i,12]<-mean(Temp[i,335:365])
}
```

```
    PrecMonth[i,12]<-mean(Prec[i,335:365])
  }
FourierBasis<-create.fourier.basis(rangeval = c(1,12),nbasis=7)
BsplineBasis<-create.bspline.basis(rangeval = c(1,12),nbasis=8)
TempMonth.fd<-Data2fd(argvals = c(1:12), y=t(TempMonth),basisobj = FourierBasis)
PrecMonth.fd<-Data2fd(argvals = c(1:12), y=t(PrecMonth),basisobj = BsplineBasis)
logitFD.pc.step(Response=StationsVars$North,FDObj=list(TempMonth.fd,PrecMonth.fd),
  nonFDvars = StationsVars[,c("altitude","longitude")])
```

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