

Package ‘FPMD’

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Type Package

Title On Functional Processes with Multiple Discontinuities

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Description This package handles the problem of estimating multiple discontinuities for a functional data process. A half-kernel approach is considered that addresses the inference of the total number, locations, and jump sizes of the changes.

License GPL (>= 2)

Depends R (>= 4.0.0)

Encoding UTF-8

LazyData true

Imports Rcpp (>= 1.0.4), RcppEigen, fdapace, zoo, locpol, pracma, extraDistr

LinkingTo Rcpp, RcppEigen

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Australian

Sydney Temperature Data.

Description

Australian daily minimum temperature climate data form year 1855 to 2012 for Sydney (Observatory Hill) station.

Usage

```
data(Australian)
```

Format

dataSS Data matrix.

Lt Observation time points.

Ly Daily minimum temperature. ...

Source

<http://www.bom.gov.au/climate/data>

Examples

```
## Not run:
data(Australian)
dataSS = Australian$dataSS
Ly = Australian$Ly
Lt = Australian$Lt

## global test
bw.seq = seq(0.01, 0.11, by = 0.02)
reject = rep(NA, length(bw.seq))
## reject H0: no jump points
for (i in 1:length(bw.seq)) {
  reject[i] = MDTest(bw = bw.seq[i], alpha = 0.05, Lt = Lt, Ly = Ly, Wtype = 'MIX')
  cat('reject:', reject[i], 'for h = ', bw.seq[i], '\n')
}

###
res = FMBreaks(Ly = Ly, Lt = Lt, wi = NULL, Wtype = "MIX", zeta = NULL,
               bw.seq = bw.seq, NbGrid = 101, kFolds = 5, refined = TRUE,
               individualCurveJump = TRUE, useBW1SE = FALSE, nRegGrid = 51,
               npoly = 1, nder = 0, alpha = 0.05, cutoff = max, M_max = 5)

## with change points
mu_jumptime <- res$mu_jumptime
mu_jumpsize <- res$mu_jumpsize
# mu_jumpsize <- res$mu_jumpsize_h_tau
h_tau <- res$h_tau
h_d <- res$h_d
zeta <- res$zeta
wi <- res$wi
```

```

mu <- res$mu
muWork <- res$muWork
obsGrid <- res$obsGrid
workGrid <- res$workGrid

###
par(mfrow = c(1, 2))
matplot(dataSS[,100:154], type = "p", pch = 1, col = 'gray', xlab = "Day",
        ylab = "Minimum Temperature", main= "", xaxt="n")
ind_x = seq.int(1, nrow(dataSS), length.out = 6)
axis(1, at=ind_x, labels=dataSS$ymd[ind_x])
lines(obsGrid*366, mu, col = "red", lwd = 1.5)
abline( v = res$mu_jumptime*366, lty = "dashed", col = "red" )
dataSS$ymd[round(res$mu_jumptime*366)]

## End(Not run)

```

avwreturn

Average value weighted returns data.

Description

The original data set consists of the daily simple returns of $n = 49$ industry portfolios from 1927 to 2020.

Usage

```
data(avwreturn)
```

Format

data_SS Data matrix.

Lt Observation time points.

Ly Daily average value weighted returns. ...

Source

http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data_library.html

Examples

```

## Not run:
data(avwreturn)
data_SS = avwreturn$data_SS
Ly = avwreturn$Ly
Lt = avwreturn$Lt

```

```

## global test
bw.seq = seq(0.01, 0.11, by = 0.01)
reject = rep(NA, length(bw.seq))
## reject H0: no jump points
for (i in 1:length(bw.seq)) {
  reject[i] = MDTest(bw = bw.seq[i], alpha = 0.05, nRegGrid = 201, Lt = Lt, Ly = Ly, Wtype = 'MIX')
  cat('reject:', reject[i], 'for h = ', bw.seq[i], '\n')
}

###
res = FPMD(Ly = Ly, Lt = Lt, wi = NULL, Wtype = "MIX", zeta = NULL,
           bw.seq = NULL, NbGrid = 101, kFolds = 5, refined = TRUE,
           individualCurveJump = TRUE, nRegGrid = 301,
           npoly = 1, nder = 0, alpha = 0.05, cutoff = max, M_max = 15)

## with change points
mu_jumptime <- res$mu_jumptime
mu_jumpsize <- res$mu_jumpsize
# mu_jumpsize <- res$mu_jumpsize_h_tau
h_tau <- res$h_tau
h_d <- res$h_d
zeta <- res$zeta
wi <- res$wi
mu <- res$mu
muWork <- res$muWork
obsGrid <- res$obsGrid
workGrid <- res$workGrid

##
par(mfrow=c(1,1))
matplot(data_SS[, -1], type = "p", pch = 1, xlab = "Dates",
        ylab = "Average Value Weighted Returns",
        main = '',
        col = rgb(0.7, 0.7, 0.7, 0.4), xaxt = "n")
ind_x = seq(1, nrow(data_SS), length.out = 18)
axis(1, at = ind_x, labels = data_SS[ind_x, 1])
abline(v = round(mu_jumptime * nrow(data_SS)), lty = 2)
lines((obsGrid * nrow(data_SS)), mu, col = 4)
data_SS$Date[round(mu_jumptime * nrow(data_SS))]

## plot mean function and confidence band
## true and estimated mean curve are based on workGrid points
plot.fmb <- function(res){

  ## confidence band
  cbandMu <- FPMD::pwCBFun(res)
  workGrid <- res$workGrid
  muWork <- res$muWork
  rho = res$rho
  tau_est = c(0, res$mu_jumptime, 1)
  mudata = t(rbind(cbandMu, muWork))

  ###
  bandGrid = t(rbind(cbandMu, workGrid))
  lband = lapply(1:(length(tau_est) - 1), function(i)

```

```

as.data.frame(bandGrid[workGrid < (tau_est[i + 1] - rho) &
               workGrid >= (tau_est[i] + rho),])

par(mfrow = c(2, 2))
ind_x = seq(1, nrow(data_SS), length.out = 18)
## 1
plot(x = NULL, y = NULL, ylim = c(-3, 3), xlim = c(0, 0.26),
      xlab = "Dates",
      ylab = "Average Value Weighted Returns",
      main = '',
      col = rgb(0.7,0.7,0.7,0.4), xaxt="n")
axis(1, at= (ind_x/nrow(data_SS))[1:5], labels=data_SS[ind_x[1:5], 1])

lapply(lband[1:5], function(x) polygon(c(x$workGrid, rev(x$workGrid)),c(x$lwCI,rev(x$upCI)),
                                       col = rgb(0.7,0.7,0.7,0.4) , border = NA))

for (i in 1:5) {
  matplot(workGrid[workGrid < (tau_est[i+1]- rho) & workGrid >= (tau_est[i]+ rho) ],
          mudata[workGrid < (tau_est[i+1] - rho) & workGrid >= (tau_est[i]+rho), ],
          type = "l", add = TRUE, lty = c(3, 3, 1), col = c(4,4,2), lwd = 2)

}
legend('top', legend = c( 'Estimated mean', 'Pointwise confidence interval'),
      cex = 1, lwd = 2, col = c(2,4), lty = c(1,3), bty = "n")
points(x = res$mu_jumptime[1:4], y = rep(-3.2, length(res$mu_jumptime[1:4])),
      pch= rep("*", length(res$mu_jumptime[1:4])), cex = 2, col = 4, xpd = TRUE)
# lines(res$obsGrid, res$mu, col = 4)
text(x = res$mu_jumptime[1:4], y = rep(-3.2, length(res$mu_jumptime[1:4])),
      labels = data_SS$Date[round(res$mu_jumptime[1:4]*nrow(data_SS))],
      xpd = TRUE, pos = 3, cex = 0.8, col = 4)

## 2
plot(x = NULL, y = NULL, ylim = c(-1.5, 1.5), xlim = c(0.26, 0.5),
      xlab = "Dates",
      ylab = "Average Value Weighted Returns",
      main = '',
      col = rgb(0.7,0.7,0.7,0.4), xaxt="n")
axis(1, at= (ind_x/nrow(data_SS))[5:8], labels=data_SS[ind_x[5:8], 1])
###
lapply(lband[5:8], function(x) polygon(c(x$workGrid, rev(x$workGrid)),c(x$lwCI,rev(x$upCI)),
                                       col = rgb(0.7,0.7,0.7,0.4) , border = NA))

for (i in 5:8) {
  matplot(workGrid[workGrid < (tau_est[i+1]- rho) & workGrid >= (tau_est[i]+ rho) ],
          mudata[workGrid < (tau_est[i+1] - rho) & workGrid >= (tau_est[i]+rho), ],
          type = "l", add = TRUE, lty = c(3, 3, 1), col = c(4,4,2), lwd = 2)

}
legend('top', legend = c( 'Estimated mean', 'Pointwise confidence interval'),
      cex = 1, lwd = 2, col = c(2,4), lty = c(1,3), bty = "n")
points(x = res$mu_jumptime[5:7], y = rep(-1.6, length(res$mu_jumptime[5:7])),
      pch= rep("*", length(res$mu_jumptime[5:7])), cex = 2, col = 4, xpd = TRUE)
# lines(res$obsGrid, res$mu, col = 4)
text(x = res$mu_jumptime[5:7], y = rep(-1.6, length(res$mu_jumptime[5:7])),
      labels = data_SS$Date[round(res$mu_jumptime[5:7]*nrow(data_SS))],
      xpd = TRUE, pos = 3, cex = 0.8, col = 4)

## 3
plot(x = NULL, y = NULL, ylim = c(-1.5, 1.5), xlim = c(0.5, 0.8),

```

```

        xlab = "Dates",
        ylab = "Average Value Weighted Returns",
        main = '',
        col = rgb(0.7,0.7,0.7,0.4), xaxt="n")
axis(1, at= (ind_x/nrow(data_SS))[9:13], labels=data_SS[ind_x[9:13], 1])
###
###
lapply(lband[8:12], function(x) polygon(c(x$workGrid, rev(x$workGrid)),c(x$lwCI,rev(x$upCI)),
                                         col = rgb(0.7,0.7,0.7,0.4) , border = NA))

for (i in 8:12) {
  matplot(workGrid[workGrid < (tau_est[i+1]- rho) & workGrid >= (tau_est[i]+ rho) ],
          mudata[workGrid < (tau_est[i+1] - rho) & workGrid >= (tau_est[i]+rho), ],
          type = "l", add = TRUE, lty = c(3, 3, 1), col = c(4,4,2), lwd = 2)
}
legend('top', legend = c( 'Estimated mean', 'Pointwise confidence interval'),
       cex = 1, lwd = 2, col = c(2,4), lty = c(1,3), bty = "n")
points(x = res$mu_jumptime[8:11], y = rep(-1.6, length(res$mu_jumptime[8:11])),
       pch= rep("*", length(res$mu_jumptime[8:11])), cex = 2, col = 4, xpd = TRUE)
# lines(res$obsGrid, res$mu, col = 4)
text(x = res$mu_jumptime[8:11], y = rep(-1.6, length(res$mu_jumptime[8:11])),
     labels = data_SS$Date[round(res$mu_jumptime[8:11]*nrow(data_SS))],
     xpd = TRUE, pos = 3, cex = 0.7, col = 4)

## 4
plot(x = NULL, y = NULL, ylim = c(-5, 5), xlim = c(0.8, 1),
     xlab = "Dates",
     ylab = "Average Value Weighted Returns",
     main = '',
     col = rgb(0.7,0.7,0.7,0.4), xaxt="n")
axis(1, at= (ind_x/nrow(data_SS))[14:18], labels=data_SS[ind_x[14:18], 1])
lapply(lband[12:14], function(x) polygon(c(x$workGrid, rev(x$workGrid)),c(x$lwCI,rev(x$upCI)),
                                         col = rgb(0.7,0.7,0.7,0.4) , border = NA))

for (i in 12:14) {
  matplot(workGrid[workGrid < (tau_est[i+1]- rho) & workGrid >= (tau_est[i]+ rho) ],
          mudata[workGrid < (tau_est[i+1] - rho) & workGrid >= (tau_est[i]+rho), ],
          type = "l", add = TRUE, lty = c(3, 3, 1), col = c(4,4,2), lwd = 2)
}
legend('topleft', legend = c( 'Estimated mean', 'Pointwise confidence interval'),
       cex = 1, lwd = 2, col = c(2,4), lty = c(1,3), bty = "n")
points(x = res$mu_jumptime[12:13], y = rep(-5.3, length(res$mu_jumptime[12:13])),
       pch= rep("*", length(res$mu_jumptime[12:13])), cex = 2, col = 4, xpd = TRUE)
# lines(res$obsGrid, res$mu, col = 4)
text(x = res$mu_jumptime[12:13], y = rep(-5.3, length(res$mu_jumptime[12:13])),
     labels = data_SS$Date[round(res$mu_jumptime[12:13]*nrow(data_SS))],
     xpd = TRUE, pos = 3, cex = 0.8, col = 4)

}

plot.fmb(res)

## individual, Xia and Qiu 2016
#### for each curve 1,5,6,16,35, 36
industry = colnames(data_SS)[-50]
par(mfrow=c(2,2))
for (i in c(5,16,35, 36)) {

```

```

plot(data_SS[, i+1],
     xlab = "Dates",
     ylab = "Average Value Weighted Returns",
     main = industry[i+1],
     col = 'gray', #rgb(0.7,0.7,0.7,0.4),
     xaxt="n",
     ylim = c(-6,6))
ind_x = seq(1, nrow(data_SS), length.out = 10)
axis(1, at= ind_x, labels = data_SS[ind_x, 1])
## individual
lines(y = res$indJump[[i]]$mu, x = Lt[[i]]*nrow(data_SS),
     col = "blue", lty = 1, lwd = 1.5)
points(x = res$indJump[[i]]$mu_jumptime*nrow(data_SS),
     y = rep(-6.4, length(res$indJump[[i]]$mu_jumptime)),
     pch= rep("*", length(res$indJump[[i]]$mu_jumptime)),
     cex = 2, col = "blue", xpd = TRUE)
text(x = res$indJump[[i]]$mu_jumptime*nrow(data_SS),
     y = rep(-6.4, length(res$indJump[[i]]$mu_jumptime)),
     labels = data_SS$Date[res$indJump[[i]]$mu_jumptime*nrow(data_SS)],
     xpd = TRUE, pos = 3, cex = 0.5, col = "blue")
## our mean
lines(obsGrid*nrow(data_SS), mu, col = "red", lty = 1, lwd = 1.5)
abline(v = round(mu_jumptime *nrow(data_SS)), col = "red", lty = "dashed")

}
par(mfrow=c(1,1))

## End(Not run)

```

Description

Multiple Thresholds detection in mean function for dense or sparse functional data. The bandwidths are selected by a k-folds cross validation.

Usage

```

FPMD(
  Ly,
  Lt,
  wi = NULL,
  Wtype = c("OBS", "SUBJ", "MIX", "OPT"),
  zeta = NULL,
  cutoff = max,
  alpha = 0.05,
  bw.seq = NULL,
  nRegGrid = 101,
  NbGrid = 101,

```

```

kernel = "epan",
npoly = 1,
nder = 0,
kFolds = 5,
refined = FALSE,
individualCurveJump = FALSE,
M_max
)

```

Arguments

Ly	A list of n vectors containing the observed values for each individual. Missing values specified by NAs are supported for dense case (dataType='dense').
Lt	A list of n vectors containing the observation time points for each individual corresponding to y. Each vector should be sorted in ascending order.
wi	A vector of weight for each observation.
Wtype	Weight structure specified for estimation. It works if wi = NULL.
zeta	Cut-off threshold for change point detection. The default is NULL.
cutoff	The method for construct the cut-off statistics. Default is max. For the case observations are much larger than subjections, we suggest to use the min.
alpha	The confidence level for the cut-off threshold. The default is alpha = 0.05, and for heavy tailed process, we suggest a higher level 0.1.
bw.seq	The selected (or user specified) bandwidth for cross validation. The default is NULL.
nRegGrid	The number of support points in each direction of covariance surface; numeric - default: 101.
NbGrid	The number of support points in the jump points detection; numeric - default: 101.
kernel	Smoothing kernel choice, common for mu and covariance; "rect", "epan", "quar" - default: "epan".
npoly	The degree of polynomial. Default is 1 for local linear smoothing.
nder	The order of derivative. Default is 0 for local linear smoothing, and should smaller than npoly.
kFolds	Number of folds for cross validation - default: 5.
refined	A refined stage for jump size estimation after the jump locations are detected; logical - default: TRUE
individualCurveJump	Jump detection for each curve by the BIC proposed in Xia and Qiu, 2016.
M_max	The maximum number of jumps for individual curve when using the BIC.

Value

A list containing the following fields: If the change points are detected, the function output the followings

mu_jumptime	The detected change point locations. If no change point detected, mu_jumptime=0
mu_jumpsize	The estimated jump sizes. If no change point detected, mu_jumpsize=0

mu_jumpsize_h_tau	The detected change point locations. If no change point detected, mu_jumpsize=0
zeta	The cut-off threshold for change point detection.
mu	For no change point case. A vector of length obsGrid containing the mean function estimate.
muWork	For change point case. A vector of length obsGrid containing the mean function estimate.
wi	The weight for each observation.
mi	Number of observations for each subject.
timings	A vector with execution times for the basic parts of the FMbreaks call.
timings_jump	A vector with execution times for the basic parts of jump detection.
workGrid	A vector of length nWorkGrid.
regGrid	A vector of length regGrid.
sigma2	Variance for measure error.
smoothCov	A nWorkGrid by nWorkGrid matrix of the smoothed covariance surface.
fittedCov	A nWorkGrid by nWorkGrid matrix of the fitted covariance surface, which is guaranteed to be non-negative definite.
optns	A list of actually used options.
h_tau	The selected (or user specified) bandwidth for jump locations estimate.
h_d	The selected (or user specified) bandwidth for jump sizes estimate.
LX	The fitted individual curves.
indJump	The jump detection for each individual curve.

References

Li, J., Li, Y., and Hsing, T. (2021). "On Functional Processes with Multiple Discontinuities".

Examples

```
## Not run:
# setting 1
## 3 change points
tau = c(.25,.5,.75) # jump locations
disc = c(.5,-.4,.4) # jump sizes
smoothmean=function(x) x^2+sin(2*pi*x) + cos(2*pi*x)
jump = function(x) sum((x>=tau)*disc) # jump function
mu_fun= function(t){smoothmean(t)+ sapply(t, jump)}
n = 400
Q = 20

# set.seed(123)
## generate gaussian process
## "cos", "sin", "fourier", "legendre01", "poly"
data = MakeFPMBData(n = n, Q = Q, proc= "gaussian", muFun = mu_fun, K = 3,
                    lambda = 1/(seq_len(3)+1)^2, sigma = 0.2, basisType = "fourier")

Lt = data$Lt
Ly = data$Ly
bw.seq = seq(0.01, 0.13, by = 0.02)
system.time(resCP <- FPMD(Ly = Ly, Lt = Lt, wi = NULL, Wtype = "MIX", zeta = NULL,
                          bw.seq = 0.06, NbGrid = 101, kFolds = 5, refined = TRUE,
```

```

individualCurveJump = FALSE, nRegGrid = 101,
npoly = 1, nder = 0, alpha = 0.05, cutoff = max))

h_tau = resCP$h_tau
h_d = resCP$h_d
zeta = resCP$zeta
mu_jumptime = resCP$mu_jumptime
mu_jumpsize = resCP$mu_jumpsize

## pointwise confidence band
PCBplot(res = resCP)
lines(resCP$obsGrid, mu_fun(resCP$obsGrid), type="l", col = 'blue' )

## End(Not run)

```

MakeFPMBData

Create a Dense Functional Data sample for a Gaussian process

Description

For a Gaussian or t process, create a dense or sparse functional data sample of size n over a [0,1] support.

Usage

```

MakeFPMBData(
  n,
  Q,
  muFun,
  rdist = runif,
  K,
  lambda,
  sigma,
  basisType = "cos",
  proc = c("gaussian", "t")
)

```

Arguments

n	number of samples to generate.
Q	an integral. The observations for per sample are following the poisson distribution with mean Q.
muFun	a function that takes a vector input and output a vector of the corresponding mean (default: zero function).
rdist	a sampler for generating the random design time points within [0, 1].
K	scalar specifying the number of basis to be used (default: 3).
lambda	vector of size K specifying the variance of each components (default: rep(1,K)).
sigma	The standard deviation of the Gaussian noise added to each observation points.

basisType	string specifying the basis type used; possible options are: 'sin', 'cos' and 'fourier' (default: 'cos') (See code of 'CreateBasis' for implementation details.)
proc	The type of random process, 'gaussian' or 't' process.

Value

A list containing the generated data with List type:

Ly	A list of n vectors containing the observed values for each individual. Missing values specified by NAs are supported for dense case (dataType='dense').
Lt	A list of n vectors containing the observation time points for each individual corresponding to y. Each vector should be sorted in ascending order.
mi	Number of observations for each subject.

Maturities	<i>Monthly U.S. Treasuries Data.</i>
------------	--------------------------------------

Description

The monthly U.S. treasuries from January 1983 to September 2010. In total, there are $n = 15$ interest rates at maturities of 3, 6, 9, 12, 18, 24, 30, 36, 48, 60, 72, 84, 96, 108, and 120 months.

Usage

```
data(Maturities)
```

Format

dmat Data matrix.

Lt Observation time points.

Ly Monthly maturity. ...

Examples

```
## Not run:
data(Maturities)
dmat = Maturities$dmat
Ly = Maturities$Ly
Lt = Maturities$Lt

## global test
bw.seq = seq(0.01, 0.11, by = 0.02)
reject = rep(NA, length(bw.seq))
## reject H0: no jump points
for (i in 1:length(bw.seq)) {
  reject[i] = MDTest(bw = bw.seq[i], alpha = 0.05, Lt = Lt, Ly = Ly, Wtype = 'MIX')
  cat('reject:', reject[i], 'for h = ', bw.seq[i], '\n')
}

##
res = FPMD(Ly = Ly, Lt = Lt, wi = NULL, Wtype = "MIX", zeta = NULL,
```

```

        bw.seq = NULL, NbGrid = 101, kFolds = 5, refined = TRUE,
        individualCurveJump = TRUE, nRegGrid = 101,
        npoly = 1, nder = 0, alpha = 0.05, cutoff = max, M_max = 10)

## with change points
mu_jumptime <- res$mu_jumptime
mu_jumpsize <- res$mu_jumpsize
# mu_jumpsize <- res$mu_jumpsize_h_tau
h_tau <- res$h_tau
h_d <- res$h_d
zeta <- res$zeta
wi <- res$wi
mu <- res$mu
muWork <- res$muWork
obsGrid <- res$obsGrid
workGrid <- res$workGrid

library(zoo)
ym <- seq(as.yearmon("1983-01"), as.yearmon("2010-09"), 1/12)
ym = format(as.Date(ym), "%Y-%m")

par(mfrow=c(1,2))
matplot(dmat, type = "p", pch = 1,
        xlab = "Dates", ylab = "Maturities(Months)",
        #main = 'Monthly U.S. Treasuries',
        col = 'gray', xaxt="n")
ind_x = c(1, round(mu_jumptime*333), 333)
axis(1, at=ind_x, labels= ym[ind_x])
abline( v = round(mu_jumptime*333), lty = "dashed", col = "red")
lines(obsGrid*333, mu, col = "red", lwd = 1.5)

## plot mean function and confidence band
## true and estimated mean curve are based on workGrid points
plot.fmb <- function(res){

  ## confidence band
  cbandMu <- FPMd::pwCBFun(res)
  workGrid <- res$workGrid
  muWork <- res$muWork
  rho = res$rho
  tau_est = c(0, res$mu_jumptime, 1)
  mudata = t(rbind(cbandMu, muWork))

  plot(x = NULL, y = NULL, ylim = c(0, 11), xlim = range(unlist(Lt)),
        xlab = "Dates",
        ylab = "Maturities(Months)",
        main = '',
        col = rgb(0.7,0.7,0.7,0.4), xaxt="n")
  ind_x = c(min(Lt[[1]]), res$mu_jumptime, max(Lt[[1]]))
  axis(1, at=ind_x, labels= ym[round(ind_x*333)])
  ###

```

```

bandGrid = t(rbind(cbandMu, workGrid))
lband = lapply(1:(length(tau_est) - 1), function(i)
  as.data.frame(bandGrid[workGrid < (tau_est[i + 1] - rho) &
    workGrid >= (tau_est[i] + rho),]))
lapply(lband, function(x) polygon(c(x$workGrid, rev(x$workGrid)),c(x$lwCI,rev(x$upCI)),
  col = rgb(0.7,0.7,0.7,0.4) , border = NA))
for (i in 1:(length(tau_est)-1)) {

  matplot(workGrid[workGrid < (tau_est[i+1]- rho) & workGrid >= (tau_est[i]+ rho) ],
    mudata[workGrid < (tau_est[i+1] - rho) & workGrid >= (tau_est[i]+rho), ],
    type = "l", add = TRUE, lty = c(3, 3, 1), col = c(3,3,2), lwd = 2)

}
legend('top', legend = c( 'Estimated mean', 'Pointwise confidence interval'),
  lwd = 2, col = c(2,3), lty = c(1,3), bty = "n")
points(x = res$mu_jumptime, y = rep(-.4, length(res$mu_jumptime)),
  pch= rep("*", length(res$mu_jumptime)), col = 4, cex =2, xpd = TRUE)
# # lines(res$obsGrid, res$mu, col = 2)
text(x = res$mu_jumptime, y = rep(-.4, length(res$mu_jumptime)),
  labels= ym[round(res$mu_jumptime*nrow(dmat))],
  xpd = TRUE, pos = 3, cex = 0.8, col = 4)

}

plot.fmb(res)

## individual
####
par(mfrow=c(3,5))
ind = c(3, 6, 9, 12, 18, 24, 30, 36, 48, 60, 72, 84, 96, 108, 120)
for (i in 1:length(ind)) {

  plot(dmat[, i],
    xlab = "Dates",
    ylab = "Maturities",
    main = paste("maturities of", ind[i], "months"),
    col = 'gray', #rgb(0.7,0.7,0.7,0.4),
    xaxt="n")
  ind_x = c(1, round(mu_jumptime*333), 333)
  axis(1, at=ind_x, labels= ym[ind_x])
  lines(y = res$indJump[[i]]$mu, x = Lt[[i]]*333,
    col = "blue", lty = 1, lwd = 1.5 )
  points(x = round(res$indJump[[i]]$mu_jumptime*333),
    y = rep(min(dmat[, i])-0.45, length(res$indJump[[i]]$mu_jumptime)),
    pch= rep("*", length(res$indJump[[i]]$mu_jumptime)),
    cex = 2, col = "blue", xpd = TRUE)
  text(x = round(res$indJump[[i]]$mu_jumptime*333),
    y = rep(min(dmat[, i])-0.45, length(res$mu_jumptime)),
    labels= ym[res$indJump[[i]]$mu_jumptime*nrow(dmat)],
    xpd = TRUE, pos = 3, cex = 0.5, col = "blue")

  ## our
  lines(obsGrid*333, mu, col = "red", lty = 1, lwd = 1.5)
  abline( v = round(mu_jumptime *nrow(dmat)), col = "red", lty = "dashed")

}
par(mfrow=c(1,1))

```

```
## End(Not run)
```

MDTest

A global test for jump points in mean function

Description

A global test for jump points in mean function

Usage

```
MDTest(
  bw,
  alpha = 0.05,
  Lt,
  Ly,
  nRegGrid = 101,
  Wtype = c("OBS", "SUBJ", "MIX", "OPT")
)
```

Arguments

bw	The selected (or user specified) bandwidth.
alpha	The confidence level for the cut-off threshold. The default is $\alpha = 0.05$, and for heavy tailed process, we suggest a higher level 0.1 .
Lt	A list of n vectors containing the observation time points for each individual corresponding to y . Each vector should be sorted in ascending order.
Ly	A list of n vectors containing the observed values for each individual. Missing values specified by NAs are supported for dense case (<code>dataType='dense'</code>).
nRegGrid	The number of support points in each direction of covariance surface; numeric - default: 101.
Wtype	Weight structure specified for estimation. It works if $w_i = \text{NULL}$.

Value

reject	1 indicates that there exists discontinuities.
--------	--

MeanBreaksFP	<i>Multiple Thresholds detection in mean function</i>
--------------	---

Description

Multiple Thresholds detection in mean function for dense or sparse functional data.

Usage

```
MeanBreaksFP(
  yin,
  xin,
  weight,
  xout,
  NbGrid,
  h_tau,
  h_d,
  zeta,
  npoly = 1,
  nder = 0,
  kernel,
  refined = FALSE
)
```

Arguments

yin	A vector containing the observed values.
xin	A vector containing the observation time points corresponding to yin.
weight	A vector of weight for each observations.
xout	A vector containing the observation time points for mean estimation.
NbGrid	The number of support points in the jump points detection; numeric - default: 101.
h_tau	The selected (or user specified) bandwidth for jump locations estimate.
h_d	The selected (or user specified) bandwidth for jump sizes estiamte.
zeta	Cut-off threshold for change point detection.
npoly	The degree of polynomial. Default is 1 for local linear smoothing.
nder	The order of derivative. Default is 0 for local linear smoothing, and should smaller than npoly.
kernel	Smoothing kernel choice, common for mu and covariance; "rect", "epan", "quar" - default: "epan".
refined	A refined stage for jump size estimation after the jump locations are detected; logical - default: TRUE

Value

A list containing the following fields: If the change points are detected, the function output the followings

<code>mu_jumptime</code>	The detected change point locations. If no change point detected, <code>mu_jumptime=0</code>
<code>mu_jumpsize</code>	The estimated jump sizes. If no change point detected, <code>mu_jumpsize=0</code>
<code>mu_jumpsize_h_tau</code>	The detected change point locations. If no change point detected, <code>mu_jumptime=0</code>
<code>mu</code>	For no change point case. A vector of length <code>obsGrid</code> containing the mean function estimate.
<code>muout</code>	For change point case. A vector of length <code>xout</code> containing the mean function estimate.
<code>xout</code>	A vector containing the observation time points for mean estimation.
<code>timings</code>	A vector with execution times for the basic parts of the <code>FMbreaks</code> call.
<code>h_tau</code>	The selected (or user specified) bandwidth for jump locations estimate.
<code>h_d</code>	The selected (or user specified) bandwidth for jump sizes estimate.

References

Li, J., Li, Y., and Hsing, T. (2021). "On Functional Processes with Multiple Discontinuities".

MMavwreturn	<i>Monthly max average value weighted returns data.</i>
-------------	---

Description

The original data set consists of the daily simple returns of $n = 49$ industry portfolios from 1927 to 2020.

Usage

```
data(MMavwreturn)
```

Format

data_SS Data matrix.

Lt Observation time points.

Ly Daily average value weighted returns. ...

Source

http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data_library.html

Examples

```
## Not run:

library(reshape2)
library(zoo)
library(dplyr)

data(MMavwreturn)
data_SS = MMavwreturn$data_SS
Ly = MMavwreturn$Ly
Lt = MMavwreturn$Lt

## global test
bw.seq = seq(0.01, 0.11, by = 0.02)
reject = rep(NA, length(bw.seq))
## reject H0: no jump points
for (i in 1:length(bw.seq)) {
  reject[i] = MDTest(bw = bw.seq[i], alpha = 0.05, Lt = Lt, Ly = Ly, Wtype = 'MIX')
  cat('reject:', reject[i], 'for h = ', bw.seq[i], '\n')
}

##
res = FPM(DLy = Ly, Lt = Lt, wi = NULL, Wtype = "MIX", zeta = NULL,
          bw.seq = NULL, NbGrid = 101, kFolds = 5, refined = TRUE,
          individualCurveJump = TRUE, nRegGrid = 501,
          npoly = 1, nder = 0, alpha = 0.05, cutoff = max, M_max = 15)

## with change points
mu_jumptime <- res$mu_jumptime
mu_jumpsize <- res$mu_jumpsize
# mu_jumpsize <- res$mu_jumpsize_h_tau
h_tau <- res$h_tau
h_d <- res$h_d
zeta <- res$zeta
wi <- res$wi
mu <- res$mu
muWork <- res$muWork
obsGrid <- res$obsGrid
workGrid <- res$workGrid

par(mfrow = c(1, 2))
matplot(data_SS[, -50], type = "p", pch = 1, xlab = "Dates",
        ylab = "Average Value Weighted Returns",
        main = '',
        col = 'gray', xaxt="n")
ind_x = seq(1, nrow(data_SS), length.out = 10)
axis(1, at=ind_x, labels=data_SS[ind_x, 50])
abline(v = round(mu_jumptime * nrow(data_SS)), lty = "dashed", col = "red")
lines(obsGrid*nrow(data_SS), mu, col = "red", lwd = 1.5)
data_SS$ym[round(mu_jumptime * nrow(data_SS))]
```

```

## plot mean function and confidence band
## true and estimated mean curve are based on workGrid points
plot.fmb <- function(res){

  ## confidence band
  cbandMu <- FPM::pwCBFun(res)
  workGrid <- res$workGrid
  muWork <- res$muWork
  rho = res$rho
  tau_est = c(0, res$mu_jumptime, 1)
  mudata = t(rbind(cbandMu, muWork))

  plot(x = NULL, y = NULL, ylim = c(0, 9), xlim = range(unlist(Lt)),
        xlab = "Dates",
        ylab = "Average Value Weighted Returns",
        main = '',
        col = rgb(0.7,0.7,0.7,0.4), xaxt="n")
  ind_x = seq(min(Lt[[1]]), max(Lt[[1]]), length.out = 10)
  axis(1, at=ind_x, labels= data_SS[ind_x*nrow(data_SS), 50])
  ###

  bandGrid = t(rbind(cbandMu, workGrid))
  lband = lapply(1:(length(tau_est) - 1), function(i)
    as.data.frame(bandGrid[workGrid < (tau_est[i + 1] - rho) &
      workGrid >= (tau_est[i] + rho),]))
  lapply(lband, function(x) polygon(c(x$workGrid, rev(x$workGrid)),c(x$lwCI,rev(x$upCI)),
    col = rgb(0.7,0.7,0.7,0.4) , border = NA))
  for (i in 1:(length(tau_est)-1)) {

    matplot(workGrid[workGrid < (tau_est[i+1]- rho) & workGrid >= (tau_est[i]+ rho) ],
      mudata[workGrid < (tau_est[i+1] - rho) & workGrid >= (tau_est[i]+rho), ],
      type = "l", add = TRUE, lty = c(3, 3, 1), col = c(4,4,2), lwd = 2)

  }
  legend('topleft', legend = c( 'Estimated mean', 'Pointwise confidence interval'),
        lwd = 2, col = c(2,4), lty = c(1,3), bty = "n")
  points(x = res$mu_jumptime, y = rep(-.35, length(res$mu_jumptime)),
        pch= rep("*", length(res$mu_jumptime)), cex = 2, col = 4, xpd = TRUE)
  # lines(res$obsGrid, res$mu, col = 4)
  text(x = res$mu_jumptime, y = rep(-.3, length(res$mu_jumptime)),
        labels = data_SS$ym[round(res$mu_jumptime*nrow(data_SS))],
        xpd = TRUE, pos = 3, cex = 0.5, col = 4)

}

plot.fmb(res)

## individual
####
industry = colnames(data_SS)[-50]
par(mfrow=c(2,2))
for (i in c(5,16,35, 36)) {
  plot(data_SS[, i],
        xlab = "Dates",
        ylab = "Average Value Weighted Returns",

```

```

        main = industry[i],
        col = 'gray', #rgb(0.7,0.7,0.7,0.4),
        xaxt="n", ylim = c(0, max(data_SS[, i])))
ind_x = seq(1, nrow(data_SS), length.out = 10)
axis(1, at=ind_x, labels=data_SS[ind_x, 50])
## individual
lines(y = res$indJump[[i]]$mu, x = Lt[[i]]*nrow(data_SS),
      col = "blue", lty = 1, lwd = 1.5 )
points(x = res$indJump[[i]]$mu_jumptime*nrow(data_SS),
       y = rep(-0.55, length(res$indJump[[i]]$mu_jumptime)),
       pch= rep("*", length(res$indJump[[i]]$mu_jumptime)),
       cex = 2, col = "blue", xpd = TRUE)
text(x = res$indJump[[i]]$mu_jumptime*nrow(data_SS),
     y = rep(-0.55, length(res$indJump[[i]]$mu_jumptime)),
     labels = data_SS$ym[res$indJump[[i]]$mu_jumptime*nrow(data_SS)],
     xpd = TRUE, pos = 3, cex = 0.5, col = "blue")
## our
lines(obsGrid*nrow(data_SS), mu, col = "red", lty = 1, lwd = 1.5)
abline( v = round(mu_jumptime *nrow(data_SS)), col = "red", lty = "dashed")
}

## End(Not run)

```

PCBplot

Plot the estimated mean function and the corresponding pointwise confidence interval

Description

Plot the estimated mean function and the corresponding pointwise confidence interval

Usage

```
PCBplot(res)
```

Arguments

`res` an FPMD class object returned by `FPMD()`.

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