Package 'chngpt'

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LazyLoad yes

LazyData yes

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Title Estimation and Hypothesis Testing for Threshold Regression

Depends R (>= 3.6)

Suggests R.rsp, RUnit, mvtnorm

Imports survival, splines, kyotil (>= 2020.10-12), boot, MASS,
methods, lme4, parallel, RhpcBLASctl

VignetteBuilder R.rsp

Description Threshold regression models are also called two-phase regression, broken-stick regression, split-point regression, structural change models, and regression kink models, with and without interaction terms. Methods for both continuous and discontinuous threshold models are included, but the support for the former is much greater. This package is described in Fong, Huang, Gilbert and Permar (2017) <DOI:10.1186/s12859-017-1863-x> and the package vignette.

License GPL (>= 2)

NeedsCompilation yes

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Contents

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chngpt   *chngpt Package*

Description

Please see the Index link below for a list of available functions. The main testing function is chngpt.test(). The main estimation function is chngptm().

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chngpt.test   *Threshold Model Hypothesis Testing*

Description

Hypothesis testing for threshold models. Only linear models and logistic models are supported at this point.

Usage

```
chngpt.test (formula.null, formula.chngpt, family=c("binomial","gaussian"), data,
    type=c("step","hinge","segmented","stegmented"),
    test.statistic=c("lr","score"), # support for score is gradually descreasing
    chngpts=NULL, lb.quantile=.1, ub.quantile=.9,
  chngpts.cnt=50, #this is set to 25 if int is weighted.two.sided or weighted.one.sided
    prec.weights=NULL,
    p.val.method=c("MC","param.boot"),
  mc.n=5e4, # 1e3 won't cut it, the p values estimated could be smaller than nominal
    boot.B=1e4,
    robust=FALSE,
    keep.fits=FALSE, verbose=FALSE
)


antoch.test (formula, data, chngpt.var, plot.=FALSE)

## S3 method for class 'chngpt.test'
plot(x, by.percentile=TRUE, both=FALSE, main=NULL, ...)
```

Arguments

formula.null	formula for the null model.
formula.chngpt	formula for the change point model. For example, suppose formula.null=y~z and we want to test whether I(x>cutff) is a significant predictor, formula.chngpt=~x If instead we are interested in testing the null that neither I(x>cutff) nor z*I(x>cutff) is a significant predictor, formula.chngpt=~x*z
data	data frame.
family	Currently only linear and logistic regression are supported.
type	step: flat before and after change point; hinge: flat before and slope after change point; segmented: slope before and after change point
test.statistic	method for testing main effects of some threshold model.
chngpts	A grid of potential change points to maximize over. If not supplied, they will be set to a vector of length chngpt.cnt equally spaced between lb.quantile and ub.quantile.
robust	Boolean.
lb.quantile	number. The lower bound in the search for change point in the unit of quantile.
ub.quantile	number. The upper bound in the search for change point in the unit of quantile.
chngpts.cnt	integer. Number of potential change points to maximize over.
mc.n	integer. Number of multivariate normal samples to generate in the Monte Carlo procedure to evaluate p-value.
verbose	Boolean.
chngpt.var	string. Name of the predictor to detect change point

`plot.`	Boolean. Whether to make a plot.
`formula`	formula.
`x`	An object of type chngpt.test.
`...`	arguments passed to or from methods
`by.percentile`	tbd
`both`	tbd
`main`	tbd
`prec.weights`	tbd
`p.val.method`	tbd
`boot.B`	tbd
`keep.fits`	tbd

Details

The model under the altnerative is the model under the null plus terms involving the threshold. For example, when the type is segmented and formula.null=~z, formula.chngpt=~x, the model under the null is ~z+x and the model under the alternative is ~z+x+(x-e)_+.

If there are missing values in the chngpt formula, those rows will be removed from the whole dataset, including null model and chngpt model.

antoch.test is only implemented for main effect only and is based on Antoch et al. (2004). Also see Fong et al. (2014).

Value

A list of class htest and chngpt.test

`p.value`	P-value
`family`	Family from input
`method`	Method from input

References

Fong, Y., Huang, Y., Gilbert, P., Permar S. (2017) chngpt: threshold regression model estimation and inference, BMC Bioinformatics, 18(1):454.

Fong Y, Di C, and Permar S. (2015) Change-Point Testing in Logistic Regression Models with Interaction Term. Statistics in Medicine. 34:1483–1494

Pastor-Barriuso, R. and Guallar, E. and Coresh, J. (2003) Transition models for change-point estimation in logistic regression. Statistics in Medicine. 22:13141

Antoch, J. and Gregoire, G. and Jaruskova, D. (2004) Detection of structural changes in generalized linear models. Statistics and probability letters. 69:315

Examples

```
dat=sim.chngpt("thresholded", "step", n=200, seed=1, beta=1, alpha=-1, x.distr="norm", e.=4,
    family="binomial")
test=chngpt.test(formula.null=y~z, formula.chngpt=~x, dat, type="step", family="binomial",
    mc.n=10)
test
plot(test)

dat=sim.chngpt("thresholded", "segmented", n=200, seed=1, beta=1, alpha=-1, x.distr="norm", e.=4,
    family="binomial")
test=chngpt.test(formula.null=y~z, formula.chngpt=~x, dat, type="segmented", family="binomial",
    mc.n=10)
test
plot(test)

test = chngpt.test (formula.null=Volume~1, formula.chngpt=~Girth, family="gaussian", data=trees,
    type="segmented", mc.n=1e4, verbose=FALSE, chngpts.cnt=100, test.statistic="lr")
test
plot(test)


## Not run:
# not run because otherwise the examples take >5s and that is a problem for R CMD check

# has interaction
test = chngpt.test(formula.null=y~z, formula.chngpt=~x*z, dat, type="step", family="binomial")
test
plot(test)


## End(Not run)
```

---

chngptm   *Threshold Models Estimation*

Description

Estimate threshold generalized linear models, Cox proportional hazards models, and linear mixed models. Supports 14 types of two-phase (one threshold) models and 1 type of three-phase (two thresholds) model.

Usage

```
chngptm (formula.1, formula.2, family, data, type = c("hinge",
 "M01", "M02", "M03", "M04", "upperhinge", "M10",
 "M20", "M30", "M40", "M21", "M12", "M21c", "M12c",
```

```
"M22", "M22c", "M31", "M13", "M33c", "segmented",
"M11", "segmented2", "M111", "step", "stegmented"),
formula.strat = NULL, weights = NULL, offset = NULL,
REML = TRUE, re.choose.by.loglik = FALSE, est.method =
c("default", "fastgrid2", "fastgrid", "grid",
"smoothapprox"), var.type = c("default", "none",
"robust", "model", "bootstrap", "all"), aux.fit =
NULL, lb.quantile = 0.05, ub.quantile = 0.95,
grid.search.max = Inf, test.inv.ci = TRUE,
boot.test.inv.ci = FALSE, bootstrap.type =
c("nonparametric", "wild", "sieve", "wildsieve",
"awb"), m.out.of.n = 0, subsampling = 0, order.max =
10, ci.bootstrap.size = 1000, alpha = 0.05, save.boot
= TRUE, b.transition = Inf, tol = 1e-04, maxit = 100,
chngpt.init = NULL, search.bound = 10, keep.best.fit =
TRUE, ncpus = 1, verbose = FALSE, ...)

chngptm.xy(x, y, type=c("step","hinge","segmented","segmented2","stegmented"),
    ...)

## S3 method for class 'chngptm'
 coef(object, ...)
## S3 method for class 'chngptm'
 residuals(object, ...)
## S3 method for class 'chngptm'
 vcov(object, var.type=NULL, ...)
## S3 method for class 'chngptm'
 print(x, ...)
## S3 method for class 'chngptm'
 predict(object, newdata = NULL,
 type = c("link", "response", "terms"), ...)
## S3 method for class 'chngptm'
 plot(x, which = NULL, xlim = NULL, ylim = NULL, lwd = 2,
         lcol = "red", lty = 1, add = FALSE, add.points = TRUE,
         add.ci = TRUE, breaks = 20, mark.chngpt = TRUE, xlab =
         NULL, ylab = NULL, plot.individual.line = FALSE, main
         = "", y.adj = NULL, auto.adj.y = FALSE, transform =
         NULL, ...)
## S3 method for class 'chngptm'
 summary(object, var.type = NULL, expo = FALSE,
 show.slope.post.threshold = FALSE, verbose = FALSE,
 boot.type = "perc", ...)
## S3 method for class 'chngptm'
 logLik(object, ...)
## S3 method for class 'chngptm'
 AIC(object, ...)
```

```
lincomb(object, comb, alpha = 0.05, boot.type = "perc")
```

Arguments

`formula.1`	The part of formula that is free of terms involving thresholded variables
`formula.2`	The part of formula that is only composed of thresholded variables
`formula.strat`	stratification formula
`family`	string. coxph or any valid argument that can be passed to glm. But variance estimate is only available for binomial and gaussian (only model-based for latter)
`data`	data frame.
`type`	type
`transform`	transform
`b.transition`	Numeric. Controls whether threshold model or smooth transition model. Default to Inf, which correponds to threshold model
`est.method`	default: estimation algorithm will be chosen optimally; fastgrid2: a super fast grid search algorithm, limited to linear regression; grid: plain grid search, works for almost all models; smoothapprox: approximates the likelihood function using a smooth function, only works for some models. fastgrid = fastgrid2, kept for backward compatibility
`var.type`	string. Different methods for estimating covariance matrix and constructing confidence intervals
`aux.fit`	a model fit object that is needed for model-robust estimation of covariance matrix
`grid.search.max`	
	The maximum number of grid points used in grid search. When doing fast grid search, grid.search.max is set to Inf internally because it does not take more time to examine all potential thresholds.
`test.inv.ci`	Boolean, whether or not to find test-inversion confidence interval for threshold
`ci.bootstrap.size`	
	integer, number of bootstrap
`alpha`	double, norminal type I error rate
`save.boot`	Boolean, whether to save bootstrap samples
`lb.quantile`	lower bound of the search range for change point estimate
`ub.quantile`	upper bound of the search range for change point estimate
`tol`	Numeric. Stopping criterion on the coefficient estimate.
`maxit`	integer. Maximum number of iterations in the outer loop of optimization.
`chngpt.init`	numeric. Initial value for the change point.
`weights`	passed to glm
`verbose`	Boolean.
`add.points`	Boolean.
`add.ci`	Boolean.

`add`	Boolean.
`breaks`	integer.
`ncpus`	Number of cores to use if the OS is not Windows.
`keep.best.fit`	Boolean.
`y`	outcome
`show.slope.post.threshold`	
	boolean
`x`	chngptm fit object.
`newdata`	newdata
`object`	chngptm fit object.
`...`	arguments passed to glm or coxph
`m.out.of.n`	sample size for m-out-of-n bootstrap, default 0 for not doing this type of bootstrap
`subsampling`	sample size for subsampling bootstrap, default 0 for not doing this type of bootstrap
`boot.test.inv.ci`	
	whether to get test inversion CI under bootstrap
`search.bound`	bounds for search for sloping parameters
`which`	an integer
`y.adj`	y.adj
`auto.adj.y`	auto.adj.y
`xlim`	xlim
`ylim`	ylim
`lwd`	lwd
`lcol`	line col
`mark.chngpt`	mark.chngpt
`xlab`	xlab
`ylab`	ylab
`offset`	offset
`lty`	lty
`boot.type`	lty
`bootstrap.type`	nonparametric: the default, classical Efron bootstrap, works for homoscedastic and heteroscedastic independent errors; sieve: works for homoscedastic autocorrelated errors; wild: works for heteroscedastic independent errors; wildsieve: works for heteroscedastic autocorrelated errors; awb: autoregressive wild bootstrap, also works for heteroscedastic autocorrelated errors, but performance may not be as good as wildsieve
`order.max`	order of autocorrelation for autocorrelated errors in sieve and wildsieve bootstrap

comb	a vector of combination coefficients that will be used to form an inner product with the estimated slope
expo	If family is binomial and expo is TRUE, coefficients summary will be shown on the scale of odds ratio instead of slopes
REML	mixed model fitting - should the estimates be chosen to optimize the REML criterion for a fixed threshold
re.choose.by.loglik	
	mixed model fitting - should the estimates be chosen to optimize likelihood (REML nor not) or goodness of fit
plot.individual.line	
	boolean
main	character string

Details

Without lb.quantile and ub.quantile, finite sample performance of estimator drops considerably! When est.method is smoothapprox, Newton-Raphson is done with initial values chosen by change point hypothesis testing. The testing procedure may be less subjective to finite sample volatility.

If var.method is bootstrap, summary of fitted model contains p values for each estimated slope. These p values are approximate p-values, obtained assuming that the bootstrap distributions are normal.

When var.method is bootstrap and the OS is not Windows, the boot package we use under the hood takes advantage of ncpus cores through parallel::mclapply.

lincomb can be used to get the estimate and CI for a linear combination of slopes.

Value

A an object of type chngptm with the following components

converged	Boolean
coefficients	vector. Estimated coefficients. The last element, named ".chngpt", is the estimated change point
test	htest. Max score test results
iter	integer. Number of iterations

References

Son, H, Fong, Y. (2020) Fast Grid Search and Bootstrap-based Inference for Continuous Two-phase Polynomial Regression Models, Environmetrics, in press.

Elder, A., Fong, Y. (2020) Estimation and Inference for Upper Hinge Regression Models, Environmental and Ecological Statistics, 26(4):287-302.

Fong, Y. (2019) Fast bootstrap confidence intervals for continuous threshold linear regression, Journal of Computational and Graphical Statistics, 28(2):466-470.

Fong, Y., Huang, Y., Gilbert, P., Permar S. (2017) chngpt: threshold regression model estimation and inference, BMC Bioinformatics, 18(1):454.

Fong, Y., Di, C., Huang, Y., Gilbert, P. (2017) Model-robust inference for continuous threshold regression models, Biometrics, 73(2):452-462.

Pastor-Barriuso, R. and Guallar, E. and Coresh, J. (2003) Transition models for change-point estimation in logistic regression. Statistics in Medicine. 22:13141

Examples

```
# also see the vignette for examples

# threshold linear regression
# for actual use, set ci.bootstrap.size to default or higher
par(mfrow=c(2,2))
types=c("hinge", "segmented", "M02", "M03")
for (type in types) {
   fit=chngptm(formula.1=logratio~1, formula.2=~range, lidar, type=type, family="gaussian",
        var.type="bootstrap", ci.bootstrap.size=100)
    print(summary(fit))
    for (i in 1:3) plot(fit, which=i)
    out=predict(fit)
    plot(lidar$range, out, main=type)
}


# with weights
dat.1=sim.chngpt("thresholded", "segmented", n=200, seed=1, beta=1, alpha=-1, x.distr="norm", e.=4,
    family="gaussian")
fit.1.a=chngptm(formula.1=y~z, formula.2=~x, family="gaussian", dat.1, type="segmented",
    est.method="fastgrid", var.type="bootstrap", weights=ifelse(dat.1$x<3.5,100,1)
    , ci.bootstrap.size=10)
summary(fit.1.a)
plot(fit.1.a)
# fit.1.a$vcov$boot.samples

## Not run:
# likelihood test, combination of slopes
dat=sim.chngpt("thresholded", "segmented", n=200, seed=1, beta=1, alpha=-1, x.distr="norm", e.=4,
    family="gaussian")
fit=chngptm(y~z, ~x, family="gaussian", dat, type="segmented", ci.bootstrap.size=100)
fit.0=lm(y~1,dat)
# likelihood ratio test using lmtest::lrtest
library(lmtest)
lrtest(fit, fit.0)
# estimate the slope after threshold using lincomb function in the chngpt package
lincomb(fit, c(0,0,1,1))

## End(Not run)


# threshold logistic regression
dat.2=sim.chngpt("thresholded", "step", n=200, seed=1, beta=1, alpha=-1, x.distr="norm", e.=4,
```

```
        family="binomial")

fit.2=chngptm(formula.1=y~z, formula.2=~x, family="binomial", dat.2, type="step", est.method="grid")
summary(fit.2)
# no variance estimates available for discontinuous threshold models such as step
# vcov(fit.2$best.fit) gives the variance estimates for the best model conditional on threshold est

# also supports cbind() formula on left hand side
set.seed(1)
dat.2$success=rbinom(nrow(dat.2), 10, 1/(1 + exp(-dat.2$eta)))
dat.2$failure=10-dat.2$success
fit.2a=chngptm(formula.1=cbind(success,failure)~z, formula.2=~x, family="binomial", dat.2,
    type="step")


# Poisson example
counts <- c(18,17,15,20,10,20,25,13,12,33,35)
x <- 1:length(counts)
print(d.AD <- data.frame(x, counts))
fit.4=chngptm(formula.1=counts ~ 1, formula.2=~x, data=d.AD, family="poisson",
    type="segmented", var.type="bootstrap", verbose=1, ci.bootstrap.size=1)
summary(fit.4)

fit.4a=chngptm(formula.1=counts ~ 1, formula.2=~x, data=d.AD, family="quasipoisson",
    type="segmented", var.type="bootstrap", verbose=1, ci.bootstrap.size=1)



## Not run:
# Not run because otherwise the examples take >5s and that is a problem for R CMD check

# coxph example
library(survival)
fit=chngptm(formula.1=Surv(time, status) ~ ph.ecog, formula.2=~age, data=lung, family="coxph",
    type="segmented", var.type="bootstrap", ci.bootstrap.size=10)
summary(fit)


# one interaction term (mtcars is part of R default installation)
# est.method will be grid as fastgrid not available for models with interaction terms yet
fit=chngptm(formula.1=mpg ~ hp, formula.2=~hp*drat, mtcars, type="segmented",
    family="gaussian", var.type="bootstrap", ci.bootstrap.size=10)
summary(fit)



# interaction, upperhinge model, bootstrap
fit=chngptm(formula.1=mpg ~ hp, formula.2=~hp*drat, mtcars, type="M10",
    family="gaussian", var.type="bootstrap", ci.bootstrap.size=10)
summary(fit)

# more than one interaction term
# subsampling bootstrap confidence interval for step model
```

```
fit=chngptm(formula.1=mpg~hp+wt, formula.2=~hp*drat+wt*drat, mtcars, type="step",
    family="gaussian", var.type="bootstrap", ci.bootstrap.size=10)
summary(fit)

# step model, subsampling bootstrap confidence intervals
fit=chngptm(formula.1=mpg~hp, formula.2=~drat, mtcars, type="step",
    family="gaussian", var.type="bootstrap", ci.bootstrap.size=10, verbose=TRUE)
summary(fit)

# higher order threshold models
dat=sim.chngpt(mean.model="thresholded", threshold.type="M22", n=500, seed=1,
    beta=c(32,2,10, 10), x.distr="norm", e.=6, b.transition=Inf, family="gaussian",
    alpha=0, sd=0, coef.z=0)
fit.0=chngptm(formula.1=y~z, formula.2=~x, dat, type="M22", family="gaussian",
    est.method="fastgrid2"); plot(fit.0)

dat=sim.chngpt(mean.model="thresholded", threshold.type="M22c", n=500, seed=1,
    beta=c(32,2,32, 10), x.distr="norm", e.=6, b.transition=Inf, family="gaussian",
    alpha=0, sd=0, coef.z=0)
fit.0=chngptm(formula.1=y~z, formula.2=~x, dat, type="M22c", family="gaussian",
    est.method="fastgrid2"); plot(fit.0)


# examples of aux.fit
fit.0=glm(yy~zz+ns(xx,df=3), data, family="binomial")
fit = chngptm (formula.1=yy~zz, formula.2=~xx, family="binomial", data, type="hinge",
    est.method="smoothapprox", var.type="all", verbose=verbose, aux.fit=fit.0,
    lb.quantile=0.1, ub.quantile=0.9, tol=1e-4, maxit=1e3)




## End(Not run)

# example of random intercept
dat=sim.twophase.ran.inte(threshold.type="segmented", n=50, seed=1)
fit = chngptm (formula.1=y~z+(1|id), formula.2=~x, family="gaussian", dat,
    type="segmented", est.method="grid", var.type="bootstrap", ci.bootstrap.size=1)
plot(fit)
out=predict(fit, re.form=NA)
plot(dat$x, out)
out.1=predict(fit, type="response", re.form=NULL)# includes re
plot(dat$x, out.1, type="p", xlab="x")
```

coef.0.ls   *Simulation Study Parameters*

Description

The true parameters used in the simulation studies.

Usage

```
data("coef.0.ls")
```

Format

The format is list of lists.

---

convert.coef predictx threshold.func
                        Helper functions

Description

Some helper functions. predictx returns confidence bands for predictions as functions of the change point variable. threshold.func returns thresholded covariates.

Usage

```
convert.coef(coef.0, threshold.type)

predictx(fit, boot.ci.type = c("perc", "basic", "symm"), alpha
 = 0.05, xx = NULL, verbose = FALSE, return.boot =
 FALSE, include.intercept = FALSE, get.simultaneous =
 TRUE)

threshold.func(threshold.type, coef, xx, x.name, include.intercept=FALSE)
```

Arguments

```
include.intercept
                coef.0
coef.0          coef.0
threshold.type  threshold.type
get.simultaneous
                threshold.type
return.boot     threshold.type
fit             fit
```

boot.ci.type	boot.ci.type
alpha	alpha
verbose	verbose
coef	coef
xx	xx
x.name	x.name

---

dat.mtct   *An Example Dataset*

Description

A dataset from the immune correlates study of Maternal To Child Transmission of HIV-1

Usage

```
data("dat.mtct")
```

Format

A data frame with 236 observations on the following 3 variables.

y a numeric vector

birth a factor with levels C-section Vaginal

NAb_SF162LS a numeric vector

References

Permar, S. R., Fong, Y., Nathan Vandergrift, Genevieve G. Fouda, Peter Gilbert, Georgia D. Tomaras, Feng Gao and Barton F. Haynes et al. (2015) Maternal HIV-1 Envelope variable loop 3-specific IgG responses and reduced risk of perinatal transmission. Journal of Clinical Investigation, 125(7):2702:2706.

---

dat.mtct.2   *An Example Dataset*

Description

A dataset from the immune correlates study of Maternal To Child Transmission of HIV-1

Usage

```
dat.mtct.2
```

Format

A data frame with 248 observations on the following 2 variables.

NAb_score a numeric vector

V3_BioV3B a numeric vector

References

Permar, S. R., Fong, Y., Nathan Vandergrift, Genevieve G. Fouda, Peter Gilbert, Georgia D. Tomaras, Feng Gao and Barton F. Haynes et al. (2015) Maternal HIV-1 Envelope variable loop 3-specific IgG responses and reduced risk of perinatal transmission. Journal of Clinical Investigation, 125(7):2702:2706.

---

double.hinge   *Fit Double Hinge Models*

Description

Fit double hinge models.

Usage

```
double.hinge(x, y, lower.y = NULL, upper.y = NULL,
        var.type = c("none", "bootstrap"), ci.bootstrap.size =
        1000, alpha = 0.05, save.boot = TRUE, ncpus = 1,
        boot.ci.type=c("percentile","symmetric"))

## S3 method for class 'double.hinge'
 plot(x, which = NULL, xlim = NULL,
 lwd = 2, lcol = "red",
 lty = 1, add.points = TRUE, add.ci = TRUE, breaks =
 20, mark.chngpt = FALSE, xlab = NULL, ylab = NULL,
 ...)
## S3 method for class 'double.hinge'
 fitted(object, ...)
## S3 method for class 'double.hinge'
 residuals(object, ...)
```

Arguments

object	x
x	x
y	y
lower.y	lower.y
upper.y	upper.y
var.type	var.type

boot.ci.type	var.type
ci.bootstrap.size	
	ci.bootstrap.size
alpha	alpha
save.boot	save.boot
ncpus	ncpus
lcol	ncpus
lwd	ncpus
which	x
xlim	x
lty	x
add.points	x
add.ci	x
breaks	x
mark.chngpt	x
xlab	x
ylab	x
...	arguments passed along

Details

If lower.y and upper.y are not supplied, min(y) is taken as the function value when x is less than or equal to the first threshold, and max(y) is taken as the function value when x is greater than or equal to the second threshold.

If the function is expected to be decreasing between the two thresholds, lower.y and upper.y should be supplied to ensure the correct fit.

mse is residual sum of squares

---

hinge.test	*A non-nested hypothesis testing problem for threshold regression models*

Description

Test a hinge effect against a linear effect

Usage

```
hinge.test(formula, cov.interest, family = c("binomial", "gaussian"), data, thres = NA,
    lb.quantile = 0.1, ub.quantile = 0.9, chngpts.cnt = 10, method = c("FDB", "B", "DB"),
    boot.B = 10000, B2 = NA, verbose = FALSE)
```

Arguments

`formula`	formula
`cov.interest`	cov.interest
`family`	family
`data`	data
`thres`	If supplied, this will be the threshold value to use in the hinge model.
`lb.quantile`	lower bound of threshold candidates in quantile
`ub.quantile`	upper bound of threshold candidates in quantile
`chngpts.cnt`	number of candidate thresholds
`method`	type of test. FDB: false double bootstrap, B: parametric bootstrap, DB: double bootstrap.
`boot.B`	number of parametric bootstrap replicates for B and FDB
`B2`	number of inner bootstrap replicates for DB
`verbose`	verbose

Value

A list of class htest

`p.value`	P-value
`chngpts`	Vector of change points evaluated
`TT`	Standardized absolute score statistics
`V.S.hat`	Estimated variance-covariance matrix of the score statistics

Author(s)

Zonglin He

References

He, Fong, Fouda, Permar. A non-nested hypothesis testing problem for threshold regression model, under review

Examples

```
dat=sim.hinge(threshold.type = 'NA',family = 'binomial',thres='NA',X.ditr = 'norm',mu.X = c(0,0,0),
    coef.X = c(0,.5,.5,.4),cov.X = diag(3),eps.sd = 1,seed = 1,n=100)
test=hinge.test(Y~X1+X2, "x", family="binomial", data=dat,'method'='FDB',boot.B=10)
test
```

---

lidar   *Light Detection and Ranging Data*

Description

LIDAR

Usage

```
data("lidar")
```

Format

A data frame with 221 observations on the following 2 variables.

range a numeric vector

logratio a numeric vector

Source

Holst, U., Hossjer, O., Bjorklund, C., Ragnarson, P. and Edner, H. (1996), Locally weighted least-squares kernel regression and statistical evaluation of LIDAR measurements, Environmetrics,7, 401-416. Wakefield (2013), Bayesian and Frequentist Regression Methods. Chapter 11 Spline and Kernel Methods.

---

nutrition   *Infant Nutrition Data*

Description

The infant nutrition dataset comprises data collected in a study on the nutrition of infants and preschool children in the north central region of the United States of America.

Usage

```
data("nutrition")
```

Format

A data frame with 72 observations on the following 2 variables.

woh weight/height ratio

age a numeric vector

Source

Eppright, E. S., Fox, H. M., Fryer, B. A., Lamkin, G. H., Vivian, V. M., Fuller, E. S. (1972). Nutrition of Infants and Preschool Children in the North Central Region of the United States of America. In World Review of Nutrition and Dietetics (Vol. 14, pp. 269-332). Karger Publishers.

---

performance.unit.test   *Perform unit testing for performance evaluation.*

Description

This function performs unit testing for performance evaulation.

Usage

```
performance.unit.test(formula.1, formula.2, family, data, B, I)
```

Arguments

formula.1	formula.1
formula.2	formula.2
family	family
data	data
B	B
I	I

---

sim.alphas   *Simulation Parameters*

Description

Simulation Parameters

Usage

```
data(sim.alphas)
```

Format

List of 6. Names: sigmoid2_norm, sigmoid2_norm3, sigmoid3_norm, sigmoid3_norm3, sigmoid4_norm, sigmoid4_norm3. Each element is a 5x4 matrix

---

sim.chngpt   *Simulation Function*

Description

Generate simulation datasets for change point Monte Carlo studies.

Usage

```
sim.chngpt (mean.model = c("thresholded", "thresholdedItxn",
 "quadratic", "quadratic2b", "cubic2b", "exp",
 "flatHyperbolic", "z2", "z2hinge", "z2segmented",
 "z2linear", "logistic"), threshold.type = c("NA",
 "M01", "M02", "M03", "M10", "M20", "M30", "M11",
 "M21", "M12", "M22", "M22c", "M31", "M13", "M33c",
 "hinge", "segmented", "upperhinge", "segmented2",
 "step", "stegmented"), b.transition = Inf, family =
 c("binomial", "gaussian"), x.distr = c("norm",
 "norm3", "norm6", "imb", "lin", "mix", "gam",
 "zbinary", "gam1", "gam2", "fixnorm", "unif"), e. =
 NULL, mu.x = 4.7, sd.x = NULL, sd = 0.3, mu.z = 0,
 alpha = NULL, alpha.candidate = NULL, coef.z =
 log(1.4), beta = NULL, beta.itxn = NULL,
 logistic.slope = 15, n, seed, weighted = FALSE,
 heteroscedastic = FALSE, ar = FALSE, verbose = FALSE)

sim.twophase.ran.inte(threshold.type, n, seed)

sim.threephase(n, seed, gamma = 1, e = 3, beta_e = 5, f = 7, beta_f = 2, coef.z = 1)
```

Arguments

threshold.type	string. Types of threshold effect to simulate, only applicable when label does not start with sigmoid.
family	string. Glm family.
n	n
mu.z	n
seed	seed
weighted	beta
beta	beta
coef.z	numeric. Coefficient for z.
beta.itxn	numeric. Coefficient for z.

`alpha`	numeric, intercept.
`mu.x`	numeric
`sd.x`	numeric
`mean.model`	numeric
`x.distr`	string. Possible values: norm (normal distribution), gam (gamma distribution). gam1 is a hack to allow e. be different
`e.`	e.
`verbose`	Boolean
`b.transition`	b.
`sd`	b.
`ar`	autocorrelation
`alpha.candidate`	
	Candidate values of alpha, used in code to determine alpha values
`e`	e
`beta_e`	beta_e
`f`	f
`beta_f`	beta_f
`logistic.slope`	beta_f
`gamma`	beta_f
`heteroscedastic`	
	Boolean.

Details

mean.model, threshold.type and b.transition all affect mean models.

Value

A data frame with following columns:

`y`	0/1 outcome
`x`	observed covariate that we are interested in
`x.star`	unobserved covariate that underlies x
`z`	additional covariate

In addition, columns starting with 'w' are covariates that we also adjust in the model; columns starting with 'x' are covariates derived from x.

Examples

```
seed=2
par(mfrow=c(2,2))
dat=sim.chngpt(mean.model="thresholded", threshold.type="hinge", family="gaussian", beta=0, n=200,
    seed=seed, alpha=-1, x.distr="norm", e.=4, heteroscedastic=FALSE)
plot(y~z, dat)
dat=sim.chngpt(mean.model="thresholded", threshold.type="hinge", family="gaussian", beta=0, n=200,
    seed=seed, alpha=-1, x.distr="norm", e.=4, heteroscedastic=TRUE)
plot(y~z, dat)
dat=sim.chngpt(mean.model="z2", threshold.type="hinge", family="gaussian", beta=1, n=200,
    seed=seed, alpha=1, x.distr="norm", e.=4, heteroscedastic=FALSE)
plot(y~z, dat)
dat=sim.chngpt(mean.model="z2", threshold.type="hinge", family="gaussian", beta=1, n=200,
    seed=seed, alpha=1, x.distr="norm", e.=4, heteroscedastic=TRUE)
plot(y~z, dat)
```

---

sim.hinge   *Simulation function*

Description

Simulate data for Monte Carlo study.

Usage

```
sim.hinge(threshold.type = c("NA", "hinge"), family = c("binomial", "gaussian"),
    thres = "NA", X.ditr = "norm", mu.X, coef.X, cov.X, eps.sd, seed, n)
```

Arguments

threshold.type	threshold.type
family	family
thres	thres
X.ditr	X.ditr
mu.X	mu.X
coef.X	coef.X
cov.X	cov.X
eps.sd	eps.sd
seed	seed
n	n

---

sim.my   *Simulate data*

Description

Simulate data

Usage

```
sim.my(n, seed, label, alpha, beta, e. = NULL, b. = NULL, tr. = NULL)
```

Arguments

n	Sample size
seed	Seed for random number generator
label	A character string which specifies the simulation scenario. sigmoid4, sigmoidgam4, elbow4
alpha	regression parameter
beta	regression parameter
e.	inflection point for the logistic transformation (the log scale)
b.	slope for the logistic transformation
tr.	threshold point

Details

When the label starts with elbow, the transformation on x.star is elbow shaped. When the label starts with sigmoid, the transformation on x.star is sigmoid shaped. Data simulated from logit(Pr(Y==1))=alpha + beta*(transformed x.star).

Value

A data frame with columns: y, x.star, x.star.expit (if label starts with sigmoid), x.star.tr (if label starts with elbow), x.bin.med (x.star dichotomized at median), x.tri (x.star trichotomized at tertiles).

Examples

```
alpha=-1; beta=log(0.2)
e.=5; b.=-30; t.=1
dat=sim.my(n=250, seed=1, label="sigmoid4", alpha, beta, e.=e., b.=b.)
```

---

sim.pastor	*Simulate data according to one of the scenarios considered in Pastor-Barriuso et al 2003*

Description

Simulate data according to one of the scenarios considered in Pastor-Barriuso et al 2003

Usage

```
sim.pastor(seed)
```

Arguments

seed   Seed for the random number generator.

Value

A data frame with columns: y, x.star, x.star.expit, and x.bin.med (x.star dichotomized at median).

Examples

```
dat=sim.pastor(seed=1)
```

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