

# Package: DoubleCone (via r-universe)

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**Type** Package

**Title** Test Against Parametric Regression Function

**Version** 1.1

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**Description** Performs hypothesis tests concerning a regression function in a least-squares model, where the null is a parametric function, and the alternative is the union of large-dimensional convex polyhedral cones. See Bodhisattva Sen and Mary C Meyer (2016) <[doi:10.1111/rssb.12178](https://doi.org/10.1111/rssb.12178)> for more details.

**License** GPL-2 | GPL-3

**Depends** graphics, grDevices, stats, utils, coneproj (>= 1.12), Matrix, MASS

**NeedsCompilation** no

**Repository** CRAN

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DoubleCone-package      *Test against a Parametric Function*

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### Description

Given a response and predictors, the null hypothesis of a parametric regression function is tested versus a large-dimensional alternative in the form of a union of polyhedral convex cones.

### Details

Package: DoubleCone  
 Type: Package  
 Version: 1.0  
 Date: 2013-10-24  
 License: GPL-2 | GPL-3

The `doubconetest` function is the generic version. The user provides an irreducible constraint matrix that defines two convex cones; the intersection of the cones is the null space of the matrix. The function provides a p-value for the test that the expected value of a vector is in the null space using the double-cone alternative.

Given a vector  $y$  and a design matrix  $X$ , the `agconst` function performs a test of the null hypothesis that the expected value of  $y$  is constant versus the alternative that it is monotone (increasing or decreasing) in each of the predictors.

The function `partlintest` performs a test of a linear model versus a partial linear model, using a double-cone alternative.

### Author(s)

Mary C Meyer and Bodhisattva Sen Maintainer: Mary C Meyer <meyer@stat.colostate.edu>

### References

TBA

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adhd      *Sub-clinical ADHD behaviors and classroom functioning in school-age children*

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### Description

Observations on children aged 9-11 in classroom settings, for a study on the effects of sub-clinical hyperactive and inattentive behaviors on social and academic functioning.

**Usage**

```
data(adhd)
```

**Format**

A data frame with 686 observations on the following 4 variables.

sex 1=boy; 2=girl

ethn 1=Colombian, 2=African American, 3=Hispanic American, 5=European American

hypb Classroom hyperactive behavior level

fcf A measure of social and academic functioning

**Source**

Brewis, A.A. Schmidt, K.L., and Meyer, M.C. (2000) ADHD-type behavior and harmful dysfunction in childhood: a cross-cultural model, *American Anthropologist*, 102(4), pp823-828.

**Examples**

```
data(adhd)
plot(adhd$hypb, adhd$fcf)
```

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agconst	<i>Test null hypothesis of constant regression function against a general, high-dimensional alternative</i>
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**Description**

Given a response and 1-3 predictors, the function will test the null hypothesis that the response and predictors are not related (i.e., regression function is constant), against the alternative that the regression function is monotone in each of the predictors. For one predictor, the alternative set is a double cone; for two predictors the alternative set is a quadruple cone, and an octuple cone alternative is used when there are three predictors.

**Usage**

```
agconst(y, xmat, nsim = 1000)
```

**Arguments**

y	A numeric response vector, length n
xmat	an n by k design matrix, full column rank, where k=1,2, or 3.
nsim	The number of data sets simulated under the null hypothesis, to estimate the null distribution of the test statistic. The default is 1000, make this larger if a more precise p-value is desired.

## Details

For one predictor, the set of non-decreasing regression functions can be described by an n-dimensional convex polyhedral cone, and the set of non-increasing regression functions is the "opposite" cone. The one-dimensional null space is the intersection of these cones. For two predictors, the alternative set consists of four cones, defined by combinations of increasing/decreasing assumptions, and for three predictors we have eight cones.

## Value

pval	The p-value for the test: H0: constant regression function
p1 through p8	monotone fits – only p1 and p2 are returned for one predictor, etc.
thetahat	The least-squares alternative fit – i.e., the projection onto the multiple-cone alternative

## Author(s)

Mary C Meyer and Bodhisattva Sen

## References

TBA

## See Also

[doubconetest](#), [partlintest](#)

## Examples

```
n=100
x1=runif(n);x2=runif(n);xmat=cbind(x1,x2)
mu=1:n;for(i in 1:n){mu[i]=20*max(x1[i]-2/3,x2[i]-2/3,0)^2}
x1g=1:21/22;x2g=x1g
par(mar=c(1,1,1,1))
y=mu+rnorm(n)
ans=agconst(y,xmat,nsim=0)
grfit=matrix(nrow=21,ncol=21)
for(i in 1:21){for(j in 1:21){
  if(sum(x1>=x1g[i]&x2>=x2g[j])>0){
    if(sum(x1<=x1g[i]&x2<=x2g[j])>0){
      f1=min(ans$thetahat[x1>=x1g[i]&x2>=x2g[j]])
      f2=max(ans$thetahat[x1<=x1g[i]&x2<=x2g[j]])
      grfit[i,j]=(f1+f2)/2
    }else{
      grfit[i,j]=min(ans$thetahat)
    }
  }else{grfit[i,j]=max(ans$thetahat)}
}}
persp(x1g,x2g,grfit,th=-50,tick="detailed",xlab="x1",ylab="x2",zlab="mu")
##to get p-value for test against constant function:
# ans=agconst(y,xmat,nsim=1000)
# ans$pval
```

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derby	<i>Kentucky Derby Winner Speed</i>
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**Description**

The Speeds of the Winning Horses in the Kentucky Derby, 1896-2012

**Usage**

```
data(derby)
```

**Format**

A data frame with 117 observations on the following 4 variables.

speed winning speed

year year of race

cond track condition with levels fast good heav mudd slop slow

name Name of the winning horse

**Source**

[www.kentuckyderby.com](http://www.kentuckyderby.com)

**Examples**

```
data(derby)
n=length(derby$year)
track=1:n*0+1
track[derby$cond=="good"]=2
track[derby$cond=="fast"]=3
plot(derby$year,derby$speed,col=track)
```

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doubconetest	<i>Test for a vector being in the null space of a double cone</i>
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**Description**

Given an  $n$ -vector  $y$  and the model  $y=m+e$ , and an  $m$  by  $n$  "irreducible" matrix  $amat$ , test the null hypothesis that the vector  $m$  is in the null space of  $amat$ .

**Usage**

```
doubconetest(y, amat, nsim = 1000)
```

**Arguments**

y	a vector of length n
amat	an m by n "irreducible" matrix
nsim	number of simulations to approximate null distribution – default is 1000, but choose more if a more precise p-value is desired

**Details**

The matrix amat defines a polyhedral convex cone of vectors  $x$  such that  $\text{amat} \% \% x \geq 0$ , and also the opposite cone  $\text{amat} \% \% x \leq 0$ . The linear space  $C$  is those  $x$  such that  $\text{amat} \% \% x = 0$ . The function provides a p-value for the null hypothesis that  $m = E(y)$  is in  $C$ , versus the alternative that it is in one of the two cones defined by amat.

**Value**

pval	The p-value for the test
p0	The least-squares fit under the null hypothesis
p1	The least-squares fit to the "positive" cone
p2	The least-squares fit to the "negative" cone

**Author(s)**

Mary C Meyer and Bodhisattva Sen

**References**

TBA, Meyer, M.C. (1999) An Extension of the Mixed Primal-Dual Bases Algorithm to the Case of More Constraints than Dimensions, *Journal of Statistical Planning and Inference*, 81, pp13-31.

**See Also**

[agconst](#), [partlintest](#)

**Examples**

```
## test against a constant function
n=100
x=1:n/n
mu=4-5*(x-1/2)^2
y=mu+rnorm(n)
amat=matrix(0,nrow=n-1,ncol=n)
for(i in 1:(n-1)){amat[i,i]=-1;amat[i,i+1]=1}
ans=doubconetest(y,amat)
ans$pval
plot(x,y,col="slategray");lines(x,mu,lty=3,col=3)
lines(x,ans$p1,col=2)
lines(x,ans$p2,col=4)
```

partlintest

*Tests linear versus partial linear model***Description**

Given a response  $y$ , a predictor  $x$ , and covariates  $z$ , the model  $y=m(x) +b'z +e$  is considered, where  $e$  is a mean-zero random error. There are three options for the null hypothesis:  $h_0=0$  tests  $m(x)$  is constant;  $h_0=1$  tests  $m(x)$  is linear, and  $h_0=2$  tests  $m(x)$  is quadratic. The (respective) alternatives are:  $m(x)$  is increasing or decreasing,  $m(x)$  is convex or concave, and  $m(x)$  is hyper-convex or hyper-concave (referring to the third derivative of  $m$ ).

**Usage**

```
partlintest(x, y, zmat, h0 = 0, nsim = 1000)
```

**Arguments**

$x$	a vector of length $n$ ; this is the main predictor of interest
$y$	a vector of length $n$ ; this is the response
$zmat$	an $n$ by $k$ matrix of covariates, should be full column rank .
$h_0$	An indicator of what null hypothesis is to be tested: $h_0=0$ for the null hypothesis: $m(x)$ is constant; $h_0=1$ tests $m(x)$ is linear, and $h_0=2$ tests $m(x)$ is quadratic.
$nsim$	The number of simulations used in creating the null distribution of the test statistic. The default is $nsim=1000$ , if a more precise p-value is desired, make $nsim$ larger.

**Details**

For the constant null hypothesis, the alternative fit is either the monotone increasing or monotone decreasing fit – whichever minimizes the sum of squared residuals. For the linear null hypothesis, the alternative fit is either convex or concave, and for the quadratic null hypothesis, the alternative fit is constrained so that the third derivative is either positive or negative over the range of  $x$ -values.

**Value**

$pval$	The p-value for the test
$p_0$	The null hypothesis fit
$p_1$	The "positive" fit
$p_2$	The "negative" fit

**Author(s)**

Mary C Meyer and Bodhisattva Sen

**References**

TBA

**See Also**[agconst,doubconetest](#)**Examples**

```

data(derby)
n=length(derby$speed)
zmat=matrix(0,nrow=n,ncol=2);zvec=1:n*0+1
zmat[derby$cond=="good",1]=1;zvec[derby$cond=="good"]=2
zmat[derby$cond=="fast",2]=1;zvec[derby$cond=="fast"]=3
ans=partlintest(derby$year,derby$speed,zmat,h0=2)
ans$pval
par(mar=c(4,4,1,1));par(mfrow=c(1,2))
plot(derby$year,derby$speed,col=zvec,pch=zvec)
points(derby$year,ans$p0,pch=20,col=zvec)
title("Null fit")
legend(1980,51.6,pch=3:1,col=3:1,legend=c("fast","good","slow"))
plot(derby$year,derby$speed,col=zvec,pch=zvec)
points(derby$year,ans$p1,pch=20,col=zvec)
title("Alternative fit")

data(adhd)
n=length(adhd$sex)
zmat=matrix(0,nrow=n,ncol=2)
zmat[adhd$sex==1,1]=1
zmat[adhd$sethn<5,2]=1
ans=partlintest(adhd$hypb,adhd$fcn,zmat,h0=1)
ans$pval
cols=c("pink3","lightskyblue3")
plot(adhd$hypb,adhd$fcn,col=cols[zmat[,1]+1],pch=zmat[,2]+1,
xlab="Hyperactive behavior level",ylab="Social and Academic Function Score")
cols2=c(2,4)
points(adhd$hypb,ans$p1,col=cols2[zmat[,1]+1],pch=20)

```



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