

# Package ‘flipscores’

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**Type** Package

**Title** Robust Testing in GLMs

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**Description** Provides robust tests for testing in GLMs,  
by sign-flipping score contributions.  
The tests are often robust against overdispersion,  
heteroscedasticity and, in some cases,  
ignored nuisance variables.  
See Hemerik and Goeman (2017) <doi:10.1007/s11749-017-0571-1>.

**License** GNU General Public License

**LazyData** TRUE

**Imports** stats, methods

**NeedsCompilation** no

**Repository** CRAN

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## R topics documented:

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flipscores

*Robust testing in GLMs***Description**

Provides two robust tests for testing in GLMs, by sign-flipping score contributions. The tests are often robust against overdispersion, heteroscedasticity and, in some cases, ignored nuisance variables.

**Usage**

```
flipscores(model0, model1, X1, alternative = "two.sided", w=1E5, scoretype="basic")
```

**Arguments**

|             |  |
|-------------|--|
| model0      | A glm or lm object representing the fit under H0.  |
| model1      | A glm or lm object representing the fit under H1. It contains one covariate in addition to those in model0. If that covariate is a factor, it should have at most 2 levels. Either model1 or X1 should be specified. |
| X1          | Vector with the values of the covariate of interest. If X1 is a factor, it should have at most 2 levels. Either model1 or X1 should be specified.  |
| alternative | Should be "greater", "less" or "two.sided"   |
| w           | The number of times that the scores are randomly sign-flipped.   |
| scoretype   | The type of score that is computed, either "basic" or "effective". Using "effective" takes into account nuisance estimation.   |

**Value**

A p-value.

**Examples**

```
set.seed(8153)
n <- 50
beta <- 2 #coefficient of interest (H0: beta=0)
gamma <- 1 #nuisance coefficient

dataset <- data.frame(x=NULL,z=NULL,y=NULL)
dataset[n,] <- NA
dataset[,1:2] <- data.frame(matrix(rnorm(n*2),n,2)) #generate covariates
colnames(dataset) <- c("x","z")
dataset$y <- rnbino(n, mu = exp(dataset[,1]*beta + dataset[,2]*gamma), size=1 )

#Y has a negative binomial distribution but we assume a Poisson model:
modelz <- glm(y~z, family=poisson,data=dataset,x=TRUE)
```

```
## Basic test:
pv1 <- flipscores(model0=modelz, X1=dataset[,1], alternative = "two.sided", scoretype="basic")
pv1 #p-value

## Test that takes into account nuisance estimation:
pv2 <- flipscores(model0=modelz, X1=dataset[,1], alternative = "two.sided", scoretype="eff")
pv2 #p-value
```

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flipscoreshd

*Robust testing in GLMs*


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### Description

Similar to the `flipscores()` function, but allows a multi-dimensional (even high-dimensional) parameter of interest.

### Usage

```
flipscoreshd(model0, X1, w=200, scoretype="basic", V="identity")
```

### Arguments

|                        |   |
|------------------------|---|
| <code>model0</code>    | A glm or lm object representing the fit under $H_0$ .   |
| <code>X1</code>        | $n$ by $d$ dataframe or matrix with the $d$ tested covariates (which have no effect under $H_0$ ).  |
| <code>w</code>         | The number of times that the scores are randomly sign-flipped.  |
| <code>scoretype</code> | The type of score that is used, either "basic" or "effective". Using "effective" takes into account nuisance estimation.  |
| <code>V</code>         | User-defined $d$ -by- $d$ matrix, where $d$ is the dimension of the parameter of interest, i.e. the nr of columns of $X1$ . One can also specify <code>V="identity"</code> to use the identity matrix, or specify <code>V="invinfo"</code> to use the inverse of the effective Fisher info. |

### Value

A p-value.

### Examples

```
set.seed(8193)
n <- 50

dataset= data.frame(matrix(rnorm(n*5),n,5))

for(k in 1:n){
```

```
dataset[k,1:5] <- dataset[k,1:5] + rnorm(1) #make the covariates correlated
}

#Y has a negative binomial distribution but we assume a Poisson model:
#Y depends only on the 4-th covariate:
dataset$y <- rnbinom(n, mu = exp(dataset[,4]*0.5), size=1 )

#We will test the hypothesis H0 that X1,X2,X3 all have coefficient 0.
#So H0 is true. X4,X5 are nuisance covariates.

formula0 <- as.formula("y ~ X4+X5")           #X4,X5: nuisance
X1 <- dataset[,1:3]                          #X1,X2,X3: tested

model0 = glm(formula0, family=poisson,na.action=na.omit,data=dataset,x=TRUE)

## Basic test (often conservative):
pv1 <- flipscoreshd(model0=model0, X1=X1, w=200, scoretype="basic",V="invinfo")
pv1 #p-value

## Test that takes into account nuisance estimation:
pv1 <- flipscoreshd(model0=model0, X1=X1, w=200, scoretype="effective",V="invinfo")
pv1 #p-value
```

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