# Package 'FABInference'

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fabtzCI

z-optimal FAB t-interval

## Description

Computation of a 1-alpha FAB t-interval using z-optimal spending function

# Usage

```
fabtzCI(y, s, dof, alpha = 0.05, psi = list(mu = 0, tau2 = 1e+05, sigma2 = 1))
```

## Arguments

у	a numeric scalar, a normally distributed statistic
S	a numeric scalar, the standard error of y
dof	positive integer, degrees of freedom for s
alpha	the type I error rate, so 1-alpha is the coverage rate
psi	a list of parameters for the spending function, including
	1. mu, the prior expectation of E[y]
	2. tau2, the prior variance of E[y]
	3. sigma2 the variance of y

#### Value

a two-dimensional vector of the left and right endpoints of the interval

## Author(s)

Peter Hoff

```
n<-10
y<-rnorm(n)
fabtzCI(mean(y),sqrt(var(y)/n),n-1)
t.test(y)$conf.int</pre>
```

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#### **Description**

Computation of a 1-alpha FAB z-interval

## Usage

```
fabzCI(y, mu, t2, s2, alpha = 0.05)
```

#### **Arguments**

У	a numeric scalar
mu	a numeric scalar
t2	a positive numeric scalar
s2	a positive numeric scalar
alpha	the type I error rate, so 1-alpha is the coverage rate

#### **Details**

A FAB interval is the "frequentist" interval procedure that is Bayes optimal: It minimizes the prior expected interval width among all interval procedures with exact 1-alpha frequentist coverage. This function computes the FAB z-interval for the mean of a normal population with an known variance, given a user-specified prior distribution determined by psi. The prior is that the population mean is normally distributed. Referring to the elements of psi as mu, t2, s2, the prior and population variance are determined as follows:

- 1. mu is the prior expectation of the mean
- 2. t2 is the prior variance of the mean
- 3. s2 is the population variance

#### Value

a two-dimensional vector of the left and right endpoints of the interval

#### Author(s)

Peter Hoff

```
y<-0
fabzCI(y,0,10,1)
fabzCI(y,0,1/10,1)
fabzCI(y,2,10,1)
fabzCI(y,0,1/10,1)
```

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glmFAB

FAB inference for generalized linear models

#### **Description**

asymptotic FAB p-values and confidence intervals for parameters in generalized linear regression models

## Usage

```
glmFAB(cformula, FABvars, lformula = NULL, alpha = 0.05,
    silent = FALSE, ...)
```

#### **Arguments**

cformula	formua for the control variables
FABvars	matrix of regressors for which to make FAB p-values and CIs
lformula	formula for the lining model (just specify right-hand side)
alpha	error rate for CIs (1-alpha CIs will be constructed)
silent	show progress (TRUE) or not (FALSE)
	additional arguments to be passed to glm

#### Value

an object of the class glmFAB which inherits from glm

#### Author(s)

Peter Hoff

```
# n observations, p FAB variables, q=2 control variables
n<-100 ; p<-25

# X is design matrix for params of interest
# beta is vector of true parameter values
# v a variable in the linking model - used to share info across betas
v<-rnorm(p) ; beta<-(2 - 2*v + rnorm(p))/3 ; X<-matrix(rnorm(n*p),n,p)/8
# control coefficients and variables
alpha1<-.5 ; alpha2<- -.5
w1<-rnorm(n)/8
w2<-rnorm(n)/8</pre>
```

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```
# simulate data
lp<-1 + alpha1*w1 + alpha2*w2 + X%*%beta
y<-rpois(n,exp(lp))
# fit model
fit<-glmFAB(y~w1+w2,X,~v,family=poisson)
fit$FABpv
fit$FABci
summary(fit) # look at p-value column</pre>
```

 ${\sf lmFAB}$ 

FAB inference for linear models

## **Description**

FAB p-values and confidence intervals for parameters in linear regression models

#### Usage

```
lmFAB(cformula, FABvars, lformula = NULL, alpha = 0.05,
  rssSplit = TRUE, silent = FALSE)
```

#### **Arguments**

cformula	formua for the control variables
FABvars	matrix of regressors for which to make FAB p-values and CIs
lformula	formula for the linking model (just specify right-hand side)
alpha	error rate for CIs (1-alpha CIs will be constructed)
rssSplit	use some residual degrees of freedom to help fit linking model (TRUE/FALSE)
silent	show progress (TRUE) or not (FALSE)

#### Value

an object of the class 1mFAB which inherits from 1m

#### Author(s)

Peter Hoff

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#### **Examples**

```
# n observations, p FAB variables, q=2 control variables
n<-100 ; p<-25
# X is design matrix for params of interest
# beta is vector of true parameter values
# v a variable in the linking model - used to share info across betas
v<-rnorm(p); beta<-(2 - 2*v + rnorm(p))/3; X<-matrix(rnorm(n*p),n,p)/8
# control coefficients and variables
alpha1<-.5; alpha2<- -.5
w1 < -rnorm(n)/8
w2 < -rnorm(n)/8
# simulate data
lp<-1 + alpha1*w1 + alpha2*w2 + X%*%beta
y<-rnorm(n,lp)
# fit model
fit < -1mFAB(y \sim w1 + w2, X, \sim v)
fit$FABpv
fit$FABci
summary(fit) # look at p-value column
```

mmleFH

Marginal MLEs for the Fay-Herriot model

## Description

Marginal MLEs for the Fay-Herriot random effects model where the covariance matrix for the sampling model is known to scale.

#### Usage

```
mmleFH(y, X, V, ss0 = 0, df0 = 0)
```

## Arguments

У	direct data following normal model $y \sim N(\theta, V\sigma^2)$
Χ	linking model predictors $\theta \sim N(X\beta, \tau^2 I)$
٧	covariance matrix to scale
ss0	prior sum of squares for estimate of $\sigma^2$
df0	prior degrees of freedom for estimate of $\sigma^2$

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#### Value

- a list of parameter estimates including
  - 1. beta, the estimated regression coefficients
  - 2. t2, the estimate of  $\tau^2$
  - 3. s2, the estimate of  $\sigma^2$

#### Author(s)

Peter Hoff

# Examples

```
n<-30 ; p<-3
X<-matrix(rnorm(n*p),n,p)
beta<-rnorm(p)
theta<-X%*%beta + rnorm(n)
V<-diag(n)
y<-theta+rnorm(n)
mmleFH(y,X,V)</pre>
```

mmleFHP

Marginal MLEs for the Fay-Herriot model with known covariance

#### **Description**

Marginal MLEs for the Fay-Herriot random effects model where the covariance matrix for the sampling model is known

### Usage

```
mmleFHP(y, X, Sigma)
```

#### **Arguments**

y direct data following normal model  $y \sim N(\theta, \Sigma)$  X linking model predictors  $\theta \sim N(X\beta, \tau^2 I)$  Sigma covariance matrix in sampling model

#### Value

- a list of parameter estimates including
  - 1. beta, the estimated regression coefficients
  - 2. t2, the estimate of  $\tau^2$

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#### Author(s)

Peter Hoff

#### **Examples**

```
n<-30 ; p<-3
X<-matrix(rnorm(n*p),n,p)
beta<-rnorm(p)
theta<-X%*%beta + rnorm(n)
Sigma<-diag(n)
y<-theta+rnorm(n)
mmleFHP(y,X,Sigma)</pre>
```

qr.lmFAB

QR decomposition

## **Description**

QR decomposition for lmFAB objects

#### Usage

```
## S3 method for class 'lmFAB' qr(x, ...)
```

#### **Arguments**

```
x lmFAB object
... see qr.lm, if you can find it
```

#### Value

qr decomposition for a design matrix

rssSplit

Residual sum of squares split

## Description

Split residual sum of squares from normal linear regression

## Usage

```
rssSplit(fit, df0 = max(1, floor(fit$df/10)), seed = -71407)
```

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### **Arguments**

fit	lm object

df0 degrees of freedom for the smaller of the two residual sums of squares

seed random seed for constructing the basis vectors of the split

#### Value

a two-dimensional vector of independent sums of squares

#### Author(s)

Peter Hoff

#### **Examples**

```
n<-30 ; p<-6 ; sigma2<-1.5
X<-matrix(rnorm(n*p),n,p)
y<-X%*%rnorm(6) + sqrt(sigma2)*rnorm(n)
ss<-rssSplit(lm(y~ -1+X))
df<-as.numeric( substring(names(ss),first=3))
ss/df</pre>
```

sfabz

Bayes-optimal spending function

#### **Description**

Compute Bayes optimal spending function

#### Usage

```
sfabz(theta, psi, alpha = 0.05)
```

#### **Arguments**

theta value of theta being tested

psi a list of parameters for the spending function, including

mu, the prior expectation of E[y]
 tau2, the prior variance of E[y]
 sigma2 the variance of y

3. sigma2 the variance of y

alpha level of test

#### **Details**

This function computes the value of s that minimizes the acceptance probability of a biased levelalpha test for a normal population with known variance, under a specified prior predictive distribution. 10 summary.glmFAB

#### Value

a scalar value giving the optimal tail-area probability

#### Author(s)

Peter Hoff

## **Examples**

```
thetas<-seq(-1,1,length=100)
s<-NULL
for(theta in thetas){ s<-c(s,sfabz(theta,list(mu=0,tau2=1,sigma2=1)) ) }
plot(thetas,s,type="l")</pre>
```

summary.glmFAB

Summarizing Generalized Linear Model Fits with FAB Inference

## Description

summary method for class glmFAB

## Usage

```
## S3 method for class 'glmFAB'
summary(object, dispersion = NULL,
    correlation = FALSE, symbolic.cor = FALSE, ...)
```

## Arguments

```
object an object of class glmFAB
dispersion see summary.glm
correlation see summary.glm
symbolic.cor see summary.glm
... see summary.glm
```

#### **Details**

A mod of summary.glm that shows FAB p-values in table

#### Value

A list of summary statistics of the fitted generalized linear model

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#### **Examples**

```
# n observations, p FAB variables, q=2 control variables
n<-100 ; p<-25
# X is design matrix for params of interest
# beta is vector of true parameter values
# v a variable in the linking model - used to share info across betas
v<-rnorm(p); beta<-(2 - 2*v + rnorm(p))/3; X<-matrix(rnorm(n*p),n,p)/8
# control coefficients and variables
alpha1<-.5 ; alpha2<- -.5
w1 < -rnorm(n)/8
w2 < -rnorm(n)/8
# simulate data
lp<-1 + alpha1*w1 + alpha2*w2 + X%*%beta
y<-rpois(n,exp(lp))
# fit model
fit<-glmFAB(y~w1+w2,X,~v,family=poisson)</pre>
fit$FABpv
fit$FABci
summary(fit) # look at p-value column
```

summary.lmFAB

Summarizing Linear Model Fits with FAB Inference

## Description

 $\hbox{summary method for class $lmFAB}\\$ 

#### Usage

```
## S3 method for class 'lmFAB'
summary(object, correlation = FALSE,
symbolic.cor = FALSE, ...)
```

# Arguments

```
object an object of class lmFAB correlation see summary.lm symbolic.cor see summary.lm see summary.lm
```

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#### **Details**

A mod of summary. 1m that shows FAB p-values in table

#### Value

A list of summary statistics of the fitted linear model

```
# n observations, p FAB variables, q=2 control variables
n<-100 ; p<-25
# X is design matrix for params of interest
# beta is vector of true parameter values
\# v a variable in the linking model – used to share info across betas
v < -rnorm(p); beta<-(2 - 2*v + rnorm(p))/3; X<-matrix(rnorm(n*p),n,p)/8
# control coefficients and variables
alpha1<-.5; alpha2<- -.5
w1 < -rnorm(n)/8
w2 < -rnorm(n)/8
# simulate data
lp<-1 + alpha1*w1 + alpha2*w2 + X%*%beta
y<-rnorm(n,lp)
# fit model
fit < -lmFAB(y \sim w1 + w2, X, \sim v)
fit$FABpv
fit$FABci
summary(fit) # look at p-value column
```

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