

Package ‘Bayesiantreg’

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Type Package

Title Bayesian t Regression for Modeling Mean and Scale Parameters

License GPL (≥ 2)

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Depends R ($\geq 4.1.0$)

Imports MASS (≥ 7.3), Matrix (≥ 1.2), mvtnorm (≥ 1.1)

Description Performs Bayesian t Regression where mean and scale parameters are modeling by linear regression structures, and the degrees of freedom parameters are estimated.

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NeedsCompilation no

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R topics documented:

Bayesiantreg-package	2
Bayesiantreg	2
BayesiantregEst	6
criteria	9
devero	10
dJ2	11
dpostb	12
dpostg	13
gammakernel	14
gammaproposal	15
glpost	16
glproposal	17
mukernel	18

muproposal	19
pJ2	20
print.summary.Bayesiantreg	21
rJ2	22
summary.Bayesiantreg	22
tabla	23
vero	24

Index	26
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Bayesiantreg-package *Function to do Bayesian t Regression: joint mean and variance modeling and estimation of the degrees of freedom*

Description

Bayesian t regression package

Details

Package: Bayesiantreg
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Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

Bayesiantreg *Bayesiantreg*

Description

Function to do Bayesian t Regression: joint mean and variance modeling and estimation of the degrees of freedom

Usage

```
Bayesiantreg(y, x, z, nsim, bini, bpri, Bpri, gini, gpri, Gpri, glini, glpri,
             type, apriori, propuesta, Maxi=NULL,
             lambda = NULL, p = NULL, burn, jump, graph1 = TRUE, graph2 = TRUE,
             graph3 = TRUE)
```

Arguments

y	object of class matrix, with the dependent variable.
x	object of class matrix, with the variables for modelling the mean.
z	object of class matrix, with the variables for modelling the precision.
nsim	a number that indicate the number of iterations.
bini	a vector with the initial values of beta.
bpri	a vector with the values of the mean of the prior of beta.
Bpri	a matrix with the values of the variance of the prior of beta.
gini	a vector with the initial values of gamma.
gpri	a vector with the values of the mean of the prior of gamma.
Gpri	a matrix with the values of the variance of the prior of gamma.
glini	a vector with the initial value of the degrees of freedom.
glpri	a vector with the value of the the prior of the degrees of freedom.
type	a vector that can take the value "D" if the prior for the degrees of freedom considered as discrete or "C" if it is continuous.
apriori	when type is "D", it is a vector that can take the values of "poi" for a Poisson prior or "unif" for a uniform prior. When type is "C", it is a vector that can take the values of "exp" for the exponential prior, "unif" for the uniform prior or "J2" for the Jeffrey's prior.
propuesta	when type is "D", it is a vector that can take the values of "poi" for a Poisson proposal, "unif" for a uniform proposal or by default the proposal made by Marin and Cepeda (␣). When type is "C", it is a vector that can take the values of "exp" for the exponential proposal, "unif" for the uniform proposal, "J2" for the Jeffrey's proposal or by default the proposal made by Marin and Cepeda (␣).
Maxi	a number indicating the maximum value for the uniform prior an the uniforme proposal.
lambda	a number indicating the mean parameter value for the Poisson prior an the Poisson proposal.
p	a number indicating the parameter value for the Jeffrey's prior an the Jeffrey's proposal.
burn	a proportion that indicate the number of iterations to be burn at the beginning of the chain.
jump	a number that indicate the distance between samples of the autocorrelated the chain, to be excluded from the final chain.
graph1	if it is TRUE present the graph of the chains without jump and burn.

graph2	if it is TRUE present the graph of the chains with jump and burn.
graph3	if it is TRUE present the graph of the standardized residuals, the the standardized residuals against the lineal predictor, the pseudo deviance residuals and the pseudo deviance residuals against the lineal predictor.

Details

The bayesian t regression allows the joint modelling of mean and variance and the estimation of the degrees of freedom of a t distributed variable, as is proposed in Marin and Cepeda (), with identical link for the mean and logarithmic for the variance, and differents discrete and continuous approach for the degrees of freedom.

Value

object of class bayesbetareg with:

coefficients	object of class matrix with the estimated coefficients of beta, gamma and degrees of freedom.
interv	object of class matrix with the estimated confidence intervals of beta, gamma and the degrees of freedom.
fitted.values	object of class matrix with the fitted values of y.
residuals	object of class matrix with the residuals of the regression.
residualsstd	object of class matrix with the standardized residuals of the regression.
residualsdev	object of class matrix with the pseudo deviance residuals of the regression.
variance	object of class matrix with the variance terms of the regression.
beta.mcmc	object of class matrix with the complete chains for beta.
gamma.mcmc	object of class matrix with the complete chains for gamma.
gl.mcmc	object of class matrix with the complete chains for the degrees of freedom.
beta.mcmc.burn	object of class matrix with the chains for beta after the burned process.
gamma.mcmc.burn	object of class matrix with the chains for gamma after the burned process.
gl.mcmc.burn	object of class matrix with the chains for the degrees of freedom after the burned process.
loglik	the logarithmic of the likelihood of the model.
AIC	AIC of the model.
BIC	BIC of the model.
DIC	BIC of the model.
PseudoDeviance	Pseudo deviance criteria of the model as is proposed by Marin and Cepeda ().
arb	acceptance percentage for beta.
arg	acceptance percentage for gamma.
argl	acceptance percentage for the degrees of freedom.
call	Call.

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.

3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished

Examples

```
n <- 10
X1 <- runif(n,0,10)
X2 <- runif(n,5,10)
X3 <- runif(n,10,15)

y1 <- c(0.09, 1.68, -2.43, 0.23, 2.94, 1.50, 3.40, 2.22, 0.28, -0.17)

betas <- c(0,0,0,0)
gammas <- c(0,0,0)
gl <- 3

x <- cbind(rep(1,n),X1,X2,X3)
z <- cbind(rep(1,n),X2,X3)
y <- y1

Bpri <- diag(rep(100,4))
bpri <- rep(0,4)

Gpri <- diag(rep(10,3))
gpri <- rep(0,3)

glpri <- 7

propuesta <- "unif2"
apriori <- "unif"
tipo <- "D"

Maxi <- 100
nsim <- 50

bini=bpri
gini=gpri
glini=glpri
```

```
reg1 <- Bayesiantreg(y, x, z, nsim=nsim, bini, bpri,
                    Bpri, gini,
                    gpri, Gpri, glini, glpri,
                    type=tipo, apriori=apriori,
                    propuesta=propuesta,
                    Maxi=Maxi, burn=0.3, jump=3,
                    graph1 = TRUE, graph2 = TRUE, graph3 = TRUE)
```

BayesiantregEst

Bayesian t regression

Description

Function to do Bayesian t Regression: joint mean and variance modeling and estimation of the degrees of freedom

Usage

```
BayesiantregEst(y, x, z, nsim, bini, bpri, Bpri, gini, gpri, Gpri, glini, glpri,
                type, apriori, propuesta, Maxi=NULL,
                lambda = NULL, p = NULL, burn, jump, graph1 = TRUE, graph2 = TRUE,
                graph3 = TRUE)
```

Arguments

y	object of class matrix, with the dependent variable.
x	object of class matrix, with the variables for modelling the mean.
z	object of class matrix, with the variables for modelling the precision.
nsim	a number that indicate the number of iterations.
bini	a vector with the initial values of beta.
bpri	a vector with the values of the mean of the prior of beta.
Bpri	a matrix with the values of the variance of the prior of beta.
gini	a vector with the initial values of gamma.
gpri	a vector with the values of the mean of the prior of gamma.
Gpri	a matrix with the values of the variance of the prior of gamma.
glini	a vector with the initial value of the degrees of freedom.
glpri	a vector with the value of the the prior of the degrees of freedom.
type	a vector that can take the value "D" if the prior for the degrees of freedom considered as discrete or "C" if it is continuous.
apriori	when type is "D", it is a vector that can take the values of "poi" for a Poisson prior or "unif" for a uniform prior. When type is "C", it is a vector that can take the values of "exp" for the exponential prior, "unif" for the uniform prior or "J2" for the Jeffrey's prior.

propuesta	when type is "D", it is a vector that can take the values of "poi" for a Poisson proposal, "unif" for a uniform proposal or by default the proposal made by Marin and Cepeda (_). When type is "C", it is a vector that can take the values of "exp" for the exponential proposal, "unif" for the uniform proposal, "J2" for the Jeffrey's proposal or by default the proposal made by Marin and Cepeda (_).
Maxi	a number indicating the maximum value for the uniform prior an the uniforme proposal.
lambda	a number indicating the mean parameter value for the Poisson prior an the Poisson proposal.
p	a number indicating the parameter value for the Jeffrey's prior an the Jeffrey's proposal.
burn	a proportion that indicate the number of iterations to be burn at the beginning of the chain.
jump	a number that indicate the distance between samples of the autocorrelated the chain, to be excluded from the final chain.
graph1	if it is TRUE present the graph of the chains without jump and burn.
graph2	if it is TRUE present the graph of the chains with jump and burn.
graph3	if it is TRUE present the graph of the standardized residuals, the the standardized residuals against the lineal predictor, the pseudo deviance residuals and the pseudo deviance residuals against the lineal predictor.

Details

The bayesian t regression allows the joint modelling of mean and variance and the estimation of the degrees of freedom of a t distributed variable, as is proposed in Marin and Cepeda ([_](#)), with identical link for the mean and logarithmic for the variance, and differents discrete and continuous approach for the degrees of freedom.

Value

object of class bayesbetareg with:

coefficients	object of class matrix with the estimated coefficients of beta, gamma and degrees of freedom.
interv	object of class matrix with the estimated confidence intervals of beta, gamma and the degrees of freedom.
fitted.values	object of class matrix with the fitted values of y.
residuals	object of class matrix with the residuals of the regression.
residualsstd	object of class matrix with the standardized residuals of the regression.
residualsdev	object of class matrix with the pseudo deviance residuals of the regression.
variance	object of class matrix with the variance terms of the regression.
beta.mcmc	object of class matrix with the complete chains for beta.
gamma.mcmc	object of class matrix with the complete chains for gamma.
gl.mcmc	object of class matrix with the complete chains for the degrees of freedom.

beta.mcmc.burn object of class matrix with the chains for beta after the burned process.
 gamma.mcmc.burn object of class matrix with the chains for gamma after the burned process.
 gl.mcmc.burn object of class matrix with the chains for the degrees of freedom after the burned process.
 AIC AIC of the model.
 BIC BIC of the model.
 DIC BIC of the model.
 PseudoDeviance a Pseudo Deviance criteria of the model as is proposed in Marin and Cepeda ().

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

Examples

```
#library(heavy)
#data(ereturns)

#y <- ereturns[,3]
#x <- cbind(rep(1,nrow(ereturns)),ereturns[,4])
#z <- x

## A priori para Beta
#Bpri <- diag(rep(100,2))
#bpri <- rep(0,2)

## A priori para Gamma
#Gpri <- diag(rep(10,2))
#gpri <- rep(0,2)

##otros parametros
#glpri <- 7

#propuesta <- "unif2"
#apriori <- "unif"
#type <- "D"

#lambda <- 0.1
```



```

#p <- 10
#Maxi <- 100

#nsim <- 100
#burn <- 0.1
#jump <- 2

#bini=bpri
#gini=gpri
#glini=glpri

#reg1 <- Bayesianreg(y, x, z, nsim, bini, bpri, Bpri, gini, gpri,Gpri, glini, glpri,
#           type, apriori, propuesta, Maxi=NULL,
#           lambda = NULL, p = NULL, burn, jump, graph1 = T, graph2 = T,
#           graph3 = T)
#summary(reg1)

```

criteria

criteria for comparison the bayesian t regression models

Description

Performs the comparison criterias for the Bayesian t Regression

Usage

```
criteria(object,...)
```

Arguments

object	object of class "Bayesianreg"
...	not used.

Details

This function allows to extract the information criteria from the model AIC, BIC, DIC and pseudo-deviance.

Value

loglik	the logarithmic of the liklihood of the model
AIC	the AiC criteria
BIC	the BIC criteria
DIC	the DIC criteria
PseudoDeviance	the pseudo deviance criteria of the model as is proposed in Marin and Cepeda ().

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>,

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

 devero

Loglikelihood for every point of the t model

Description

Calculate the loglikelihood for every point of the t model

Usage

```
devero(y, mu, sigma2, grados)
```

Arguments

y	object of class matrix, with the dependent variables.
mu	object of class matrix, with the mean of the model.
sigma2	object of class matrix, with the variace of the model.
grados	a vector with the degrees of freedom of the model.

Details

Calculate the loglikelihood for the t model as proposed by Marin and Cepeda ().

Value

1 a value with the loglikelihood for the t model

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

dJ2

density of the Jeffrey's distribution

Description

calculates the density of the Jeffrey's distribution

Usage

`dJ2(gl.ini, p)`

Arguments

<code>gl.ini</code>	a vector with the number to evaluate in the density.
<code>p</code>	a number indicating the parameter value for the Jeffrey's prior an the Jeffrey's proposal

Details

Calculates the density of the Jeffrey's distribution

Value

`J1` the value of the density

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

dpostb

Posterior value of beta

Description

Calculate a value for posterior density for beta parameter

Usage

```
dpostb(y, x, z, betas, gammas, gl, bpri, Bpri)
```

Arguments

y	object of class matrix, with the dependen variables.
x	object of class matrix, with the variables for modelling the mean.
z	object of class matrix, with the variables for modelling the variance.
betas	a vector with the proposal beta parameters.
gammas	a vector with the proposal gamma parameters.
gl	a vector with the proposal degrees of freedom parameters.
bpri	a vector with the values of the mean of the prior of beta.
Bpri	a matrix with the values of the variance of the prior of beta.

Details

Generate the posterior density for the beta proposed by Marin and Cepeda ().

Value

value a value with the posterior denity for beta

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

dpostg *Posterior density of gamma*

Description

Propose a value for posterior density of the gamma parameter

Usage

```
dpostg(y, x, z, betas, gammas, gl, gpri, Gpri)
```

Arguments

y	object of class matrix, with the dependent variables.
x	object of class matrix, with the variables for modelling the mean.
z	object of class matrix, with the variables for modelling the variance.
betas	a vector with the proposal beta parameters.
gammas	a vector with the proposal gamma parameters.
gl	a vector with the proposal degrees of freedom parameter.
gpri	a vector with the values of the mean of the prior of gamma.
Gpri	a matrix with the values of the variance of the prior of gamma.

Details

Generate the posterior density for the gamma proposed by Marin and Cepeda ().

Value

value a value with the posterior density for gamma

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

gammakernel *the probability of a gamma parameter from the probability density function defined by old parameters.*

Description

evaluate the probability of a gamma parameter from the probability density function defined by old parameters.

Usage

```
gammakernel(y, x, z, betas.ini, gammas.now, gammas.old, gl.ini, gpri, Gpri)
```

Arguments

y	object of class matrix, with the dependent variable
x	object of class matrix, with the variables for modelling the mean
z	object of class matrix, with the variables for modelling the variance
betas.ini	a vector with the beta parameters that define the old p.d.f
gammas.now	a vector with the gamma parameters - new parameters - to evaluate in the old p.d.f
gammas.old	a vector with the gamma parameters that define the old p.d.f
gl.ini	a vector with the degrees of freedom parameters that define the old p.d.f
gpri	a vector with the initial values of gamma
Gpri	a matrix with the initial values of the variance of gamma

Details

Evaluate the probability of a gamma parameter from the probability density function defined by old parameters, according with the model proposed by Marin and Cepeda-Cuervo (␣).

Value

value a vector with the probability for the gamma parameter from the probability density function defined by old parameters.

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

 gammaproposal

A proposal for gamma parameters

Description

Propose a value for the gamma parameters

Usage

```
gammaproposal(y, x, z, betas.ini, gammas.ini, gl.ini, gpri, Gpri)
```

Arguments

y	object of class matrix, with the dependent variable
x	object of class matrix, with the variables for modelling the mean
z	object of class matrix, with the variables for modelling the variance
betas.ini	a vector with the previous proposal beta parameters
gammas.ini	a vector with the previous proposal gamma parameters
gl.ini	a vector with the previous proposal degrees of freedom parameter
gpri	a vector with the values of the mean of the prior of gamma.
Gpri	a matrix with the values of the variance of the prior of gamma.

Details

Generate a proposal for the gamma parameters according to the model proposed by Marin and Cepeda-Cuervo ().

Value

gammas.pro a number with the proposal for the gamma parameters.

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

glpost

Posterior value of the degrees of freedom

Description

Calculate a value for posterior density of the degrees of freedom parameter

Usage

```
glpost(y, x, z, betas.ini, gammas.ini, gl.ini, Maxi, lambda, p, prior, type)
```

Arguments

y	object of class matrix, with the dependent variables.
x	object of class matrix, with the variables for modelling the mean.
z	object of class matrix, with the variables for modelling the variance.
betas.ini	a vector with the proposal beta parameters.
gammas.ini	a vector with the proposal gamma parameters.
gl.ini	a vector with the proposal degrees of freedom parameter.
Maxi	a number indicating the maximum value for the uniform prior an the uniforme proposal
lambda	a number indicating the mean parameter value for the Poisson prior an the Poisson proposal
p	a number indicating the parameter value for the Jeffrey's prior an the Jeffrey's proposal
type	a vector that can take the value "D" if the prior for the degrees of freedom considered as discrete or "C" if it is continuous.
prior	when type is "D", it is a vector that can take the values of "poi" for a Poisson prior or "unif" for a uniform prior. When type is "C", it is a vector that can take the values of "exp" for the exponential prior, "unif" for the uniform prior or "J2" for the Jeffrey's prior.

Details

Generate the posterior density for the degrees of freedom proposed by Marin and Cepeda ().

Value

value a value with the posterior density for the degrees of freedom

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

glproposal

A proposal for degrees of freedom parameter

Description

Propose a value for the degrees of freedom parameter

Usage

```
glproposal(gl.ini, lambda, p, Maxi, matriz, propuesta, type)
```

Arguments

gl.ini	a vector with the previous proposal degrees of freedom parameter
lambda	a number indicating the mean parameter value for the Poisson prior an the Poisson proposal.
p	a number indicating the parameter value for the Jeffrey's prior an the Jeffrey's proposal.
Maxi	a number indicating the maximum value for the uniform prior an the uniforme proposal.
matriz	a matrix generate by the function tabla of the bayesiantreg package.
propuesta	when type is "D", it is a vector that can take the values of "poi" for a Poisson proposal, "unif" for a uniform proposal or by default the proposal made by Marin and Cepeda (). When type is "C", it is a vector that can take the values of "exp" for the exponential proposal, "unif" for the uniform proposal, "J2" for the Jeffrey's proposal or by default the proposal made by Marin and Cepeda ().
type	a vector that can take the value "D" if the prior for the degrees of freedom considered as discrete or "C" if it is continuous.

Details

Generate a proposal for the gamma parameter according to the model proposed by Marin and Cepeda-Cuervo (□).

Value

gl.pro a number with the proposal for the degrees of freedom parameter.

Author(s)

Margarita Marin <mmarinj@una1.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@una1.edu.co>

References

1. Marin and Cepeda-Cuervo (□). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

mukernel

the probability of a beta parameter from the probability density function defined by old parameters

Description

evaluate the probability of a beta parameter from the probability density function defined by old parameters

Usage

mukernel(y, x, z, betas.now, betas.old, gammas.ini, gl.ini, bpri, Bpri)

Arguments

y	object of class matrix, with the dependent variable
x	object of class matrix, with the variables for modelling the mean
z	object of class matrix, with the variables for modelling the variance
betas.now	a vector with the beta parameters - new parameters - to evaluate in the old p.d.f
betas.old	a vector with the beta parameters that define the old p.d.f
gammas.ini	a vector with the gammas parameters that define the old p.d.f
gl.ini	a vector with the degrees of freedom parameter that define the old p.d.f
bpri	a vector with the initial values of beta
Bpri	a matrix with the initial values of the variance of beta

Details

Evaluate the probability of a beta parameter from the probability density function defined by old parameters, according with the model proposed by Cepeda(2001) and Cepeda and Gamerman(2005).

Value

value a matrix with the probability for the beta parameter from the probability density function defined by old parameters

Author(s)

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References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

muproposal

A proposal for beta parameter

Description

Propose a value for the beta parameter

Usage

```
muproposal(y, x, z, betas.ini, gammas.ini, gl.ini, bpri, Bpri)
```

Arguments

y	object of class matrix, with the dependent variable
x	object of class matrix, with the variables for modelling the mean
z	object of class matrix, with the variables for modelling the variance
betas.ini	a vector with the previous proposal beta parameters
gammas.ini	a vector with the previous proposal gamma parameters
gl.ini	a vector with the previous proposal degrees of freedom parameter
bpri	a vector with the values of the mean of the prior of beta.
Bpri	a matrix with the values of the variance of the prior of beta.

Details

Generate a proposal for the beta parameters according to the model proposed by Marin and Cepeda-Cuervo ().

Value

betas.pro a number with the proposal for the beta parameters.

Author(s)

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References

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3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

pJ2

density of the Jeffrey's distribution

Description

calculates the probability of the Jeffrey's distribution

Usage

pJ2(gl.ini, p)

Arguments

gl.ini	a vector with the number to evaluate in the density.
p	a number indicating the parameter value for the Jeffrey's prior an the Jeffrey's proposal

Details

Calculates the probability of the Jeffrey's distribution

Value

J1I the value of the probability

Author(s)

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```
print.summary.Bayesianreg
    print the summary of the Bayesian t regression
```

Description

Print the summary BBayesian t Regression: joint mean and variance modeling and estimation of the degrees of freedom

Usage

```
## S3 method for class 'summary.Bayesianreg'
print(x, ...)
```

Arguments

x	object of class Bayesianreg
...	not used.

Value

Print the summary Bayesian t Regression: joint mean and variance modeling and estimation of the degrees of freedom

Author(s)

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References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. Brazilian Journal of Probability and Statistics. 14, 207-221

rJ2 *random number from the Jeffrey's distribution*

Description

generates random numbers from the Jeffrey's distribution

Usage

```
rJ2(n, matriz, min, max)
```

Arguments

n	a number that indicates the number of random values that will be generated from the Jeffrey's distribution.
matriz	a matrix generated by the function tabla of the bayesianreg package.
min	a number indicating the minimum number that can be generated from the Jeffrey's distribution.
max	a number indicating the maximum number that can be generated from the Jeffrey's distribution.

Details

generates random numbers from the Jeffrey's distribution

Value

grados the random number

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

summary.Bayesianreg *summary of the Bayesian t regression*

Description

Summarized the Bayesian Bayesian t Regression: joint mean and variance modeling and estimation of the degrees of freedom

Usage

```
## S3 method for class 'Bayesianreg'
summary(object, ...)
```


Details

generates a table with diferente probabilities and associated numbers from the Jeffrey's distribution

Value

matriz a matrix with the generated probabilities and associated numbers from the Jeffrey's distribution

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

vero

Loglikelihood for the t model

Description

Calculate the loglikelihood for the t model

Usage

```
vero(y, mu, sigma2, grados)
```

Arguments

y object of class matrix, with the dependent variables.
mu object of class matrix, with the mean of the model.
sigma2 object of class matrix, with the variace of the model.
grados a vector with the degrees of freedom of the model.

Details

Calculate the loglikelihood for the t model as proposed by Marin and Cepeda ().

Value

l a value with the loglikelihood for the t model

Author(s)

Margarita Marin <mmarinj@unal.edu.co>, Edilberto Cepeda-Cuervo <ecepdac@unal.edu.co>

References

1. Marin and Cepeda-Cuervo (). A Bayesian regression model for the non-standardized t distribution with location, scale and degrees of freedom parameters. Unpublished
2. Cepeda-Cuervo E. (2001). Modelagem da variabilidade em modelos lineares generalizados. Unpublished Ph.D. tesis. Instituto de Matematicas. Universidade Federal do Rio do Janeiro.
3. Cepeda C., E. and Gamerman D. (2001). Bayesian Modeling of Variance Heterogeneity in Normal Regression Models. *Brazilian Journal of Probability and Statistics*. 14, 207-221

Index

* Bayesian

Bayesiantreg, 2
BayesiantregEst, 6
criteria, 9
devero, 10
dJ2, 11
dpostb, 12
dpostg, 13
gammakernel, 14
gammaproposal, 15
glpost, 16
glproposal, 17
mukernel, 18
muproposal, 19
pJ2, 20
print.summary.Bayesiantreg, 21
rJ2, 22
summary.Bayesiantreg, 22
tabla, 23
vero, 24

* Degrees of Freedom

Bayesiantreg, 2
BayesiantregEst, 6
criteria, 9
devero, 10
dJ2, 11
dpostb, 12
dpostg, 13
gammakernel, 14
gammaproposal, 15
glpost, 16
glproposal, 17
mukernel, 18
muproposal, 19
pJ2, 20
print.summary.Bayesiantreg, 21
rJ2, 22
summary.Bayesiantreg, 22
tabla, 23

vero, 24

* Meancovariance modelling

Bayesiantreg, 2
BayesiantregEst, 6
criteria, 9
devero, 10
dJ2, 11
dpostb, 12
dpostg, 13
gammakernel, 14
gammaproposal, 15
glpost, 16
glproposal, 17
mukernel, 18
muproposal, 19
pJ2, 20
print.summary.Bayesiantreg, 21
rJ2, 22
summary.Bayesiantreg, 22
tabla, 23
vero, 24

* Metropolis Hastings

Bayesiantreg, 2
BayesiantregEst, 6
criteria, 9
devero, 10
dJ2, 11
dpostb, 12
dpostg, 13
gammakernel, 14
gammaproposal, 15
glpost, 16
glproposal, 17
mukernel, 18
muproposal, 19
pJ2, 20
print.summary.Bayesiantreg, 21
rJ2, 22
summary.Bayesiantreg, 22

tabla, 23
 vero, 24
* **package**
 Bayesiantreg-package, 2
* **t**
 Bayesiantreg, 2
 BayesiantregEst, 6
 criteria, 9
 devero, 10
 dJ2, 11
 dpostb, 12
 dpostg, 13
 gammakernel, 14
 gammaproposal, 15
 glpost, 16
 glproposal, 17
 mukernel, 18
 muproposal, 19
 pJ2, 20
 print.summary.Bayesiantreg, 21
 rJ2, 22
 summary.Bayesiantreg, 22
 tabla, 23
 vero, 24

Bayesiantreg, 2
Bayesiantreg-package, 2
BayesiantregEst, 6

criteria, 9

devero, 10
dJ2, 11
dpostb, 12
dpostg, 13

gammakernel, 14
gammaproposal, 15
glpost, 16
glproposal, 17

mukernel, 18
muproposal, 19

pJ2, 20
print.summary.Bayesiantreg, 21

rJ2, 22

summary.Bayesiantreg, 22

 tabla, 23
 vero, 24