

# Package ‘AdjBQR’

October 30, 2016

**Type** Package

**Title** Adjusted Bayesian Quantile Regression Inference

**Version** 1.0

**Date** 2016-10-25

**Author** Huixia Judy Wang and Yunwen Yang

**Maintainer** Huixia Judy Wang <judywang@gwu.edu>

**Description** Adjusted inference for Bayesian quantile regression based on asymmetric Laplace working likelihood, for details see Yang, Y., Wang, H. and He, X. (2015), Posterior inference in Bayesian quantile regression with asymmetric Laplace likelihood, International Statistical Review, 2015 <doi:10.1111/insr.12114>.

**Depends** quantreg, MHadaptive, coda, survival

**License** GPL-3

**RoxygenNote** 5.0.1

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2016-10-30 16:58:03

## R topics documented:

BCQR . . . . .	2
BQR . . . . .	3
li_powell . . . . .	4
li_reg . . . . .	5
sigma_est . . . . .	6
sigma_est_powell . . . . .	7

**Index**

8

BCQR

*Adjusted Bayesian Censored Quantile Regression***Description**

Bayesian quantile regression based on asymmetric-Laplace-type likelihood with posterior variance adjustment

**Usage**

```
BCQR(y, x, tau = 20000, burn_in = 4000, prop_cov = NULL,
      level = 0.9)
```

**Arguments**

y	the observed response vector that is left censored at zero
x	the design matrix. If the first column of x is not all ones, a column of ones will be added.
tau	the quantile level of interest
niter	integer: number of iterations to run the chain for. Default 20000.
burn_in	integer: discard the first burn_in values. Default 100.
prop_cov	covariance matrix giving the covariance of the proposal distribution. This matrix need not be positive definite. If the covariance structure of the target distribution is known (approximately), it can be given here. If not given, the diagonal will be estimated via the Fisher information matrix.
level	nominal confidence level for the credible interval

**Details**

The function returns the unadjusted and adjusted posterior standard deviation, and unadjusted and adjusted credible intervals for Bayesian censored quantile regression based on asymmetric-Laplace-type working likelihood. The asymmetric-Laplace-type likelihood is based on the objective function of the Powell's estimator in Powell (1986).

**Value**

A list of the following components is returned

- estpar: posterior mean of the regression coefficient vector
- PSD: posterior standard deviation without adjustment
- PSD.adj: posterior standard deviation with adjustment
- CI.BAL: credible interval without adjustment
- CI.BAL.adj: credible interval with adjustment
- sig: estimated scale parameter
- MCMCszie: effective size of the chain

## References

- Powell, J. L. (1986). Censored regression quantiles. *Journal of Econometrics*, 32, 143-155.
- Yang, Y., Wang, H. and He, X. (2015). Posterior inference in Bayesian quantile regression with asymmetric Laplace likelihood. *International Statistical Review*, 2015. doi: 10.1111/insr.12114.

## Examples

```
#A simulation example
library(AdjBQR)
n=200
set.seed(12368819)
x1=rnorm(n)
x2=rnorm(n)
ystar=3/4+2*x1+3*x2+rt(n,df=3)
y=ystar*(ystar>0)
delta=1*(ystar>0)
x = cbind(x1, x2)
## Bayesian censored quantile regression based on asymmetric-Laplace-type likelihood
BCQR(y, x, tau=0.5, level=0.9)
```

BQR

*Adjusted Bayesian Quantile Regression*

## Description

Bayesian quantile regression based on asymmetric Laplace likelihood with posterior variance adjustment

## Usage

```
BQR(y, x, tau, niter = 20000, burn_in = 4000, prop_cov = NULL,
level = 0.9)
```

## Arguments

y	the response vector
x	the design matrix. If the first column of x is not all ones, a column of ones will be added.
tau	the quantile level of interest
niter	integer: number of iterations to run the chain for. Default 20000.
burn_in	integer: discard the first burn_in values. Default 100.
prop_cov	covariance matrix giving the covariance of the proposal distribution. This matrix need not be positive definite. If the covariance structure of the target distribution is known (approximately), it can be given here. If not given, the diagonal will be estimated via the Fisher information matrix.
level	nominal confidence level for the credible interval

## Details

The function returns the unadjusted and adjusted posterior standard deviation, and unadjusted and adjusted credible intervals for Bayesian quantile regression based on asymmetric Laplace working likelihood.

## Value

A list of the following components is returned  
 estpar: posterior mean of the regression coefficient vector  
 PSD: posterior standard deviation without adjustment  
 PSD.adj: posterior standard deviation with adjustment  
 CI.BAL: credible interval without adjustment  
 CI.BAL.adj: credible interval with adjustment  
 sig: estimated scale parameter  
 MCMCszie: effective size of the chain

## References

Yang, Y., Wang, H. and He, X. (2015). Posterior inference in Bayesian quantile regression with asymmetric Laplace likelihood. International Statistical Review, 2015. doi: 10.1111/insr.12114.

## Examples

```
#A simulation example
library(AdjBQR)
n=200
set.seed(12368819)
x1 = rnorm(n)
x2 = rnorm(n)
y=2*x1+2*x2+rt(n,df=3)
x = cbind(1, x1, x2)
## Bayesian quantile regression based on asymmetric Laplace likelihood
BQR(y, x, tau=0.5, level=0.9)
```

**li\_powell**

*Asymmetric-Laplace-type Working Likelihood For Censored Quantile Regression*

## Description

Asymmetric-Laplace-type working likelihood for linear quantile regression with responses subject to left censoring at zero

## Usage

```
li_powell(pars, y, x, tau, sig)
```

### Arguments

pars	regression coefficient vector
y	the response vector
x	the design matrix with one in the first column corresponding to the intercept
tau	the quantile level
sig	scale parameter sigma

### Details

The asymmetric-Laplace-type working likelihood is proportional to exponential of the negative Powell objective function for censored quantile regression

### Value

the working log (asymmetric Laplace-type) likelihood function (the part involving the regression coefficients)

### References

- Powell, J. L. (1986). Censored regression quantiles. *Journal of Econometrics*, 32, 143-155.
- Yang, Y., Wang, H. and He, X. (2015). Posterior inference in Bayesian quantile regression with asymmetric Laplace likelihood. *International Statistical Review*, 2015. doi: 10.1111/insr.12114.

### li\_reg

*Asymmetric Laplace Working Likelihood For Linear Quantile Regression*

### Description

Asymmetric Laplace Working Likelihood For Linear Quantile Regression

### Usage

```
li_reg(pars, y, x, tau, sig)
```

### Arguments

pars	regression coefficient vector
y	the response vector
x	the design matrix with one in the first column corresponding to the intercept
tau	the quantile level
sig	scale parameter sigma

## Details

The asymmetric Laplace working likelihood is proportional to exponential of the negative quantile objective function for linear quantile regression

## Value

the working log (asymmetric Laplace) likelihood function (the part involving the regression coefficients)

## References

Yang, Y., Wang, H. and He, X. (2015). Posterior inference in Bayesian quantile regression with asymmetric Laplace likelihood. International Statistical Review, 2015. doi: 10.1111/insr.12114.

*sigma\_est*

*Estimation of the Scale Parameter for Quantile Regression*

## Description

Estimation of the Scale Parameter for Quantile Regression

## Usage

```
sigma_est(y, x)
```

## Arguments

- |          |                                                                               |
|----------|-------------------------------------------------------------------------------|
| <i>y</i> | the response vector                                                           |
| <i>x</i> | the design matrix with one in the first column corresponding to the intercept |

## Value

the estimated scale parameter sigma

---

sigma\_est\_powell      *Estimation of the Scale Parameter for Censored Quantile Regression*

---

**Description**

Estimation of the Scale Parameter for Censored Quantile Regression

**Usage**

```
sigma_est_powell(y, x)
```

**Arguments**

y	the response vector
x	the design matrix with one in the first column corresponding to the intercept

**Value**

the estimated scale parameter sigma

# Index

BCQR, [2](#)

BQR, [3](#)

li\_powell, [4](#)

li\_reg, [5](#)

sigma\_est, [6](#)

sigma\_est\_powell, [7](#)