

# Package: NBtsVarSel (via r-universe)

October 10, 2024

**Type** Package

**Title** Variable Selection in a Specific Regression Time Series of Counts

**Version** 1.0

**Date** 2023-07-17

**Description** Performs variable selection in sparse negative binomial GLARMA (Generalised Linear Autoregressive Moving Average) models. For further details we refer the reader to the paper Gomtsyan (2023), <[arXiv:2307.00929](https://arxiv.org/abs/2307.00929)>.

**License** GPL-2

**Depends** R (>= 3.5.0), Matrix, glmnet, stats, MASS, mpath, ggplot2

**VignetteBuilder** knitr

**Suggests** knitr, markdown, formatR

**NeedsCompilation** no

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**Repository** <https://mgomtsyan.r-universe.dev>

**RemoteUrl** <https://github.com/cran/NBtsVarSel>

**RemoteRef** HEAD

**RemoteSha** 633b933f665ae113a58212810e5119adcf38f455

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## Description

NBtsVarSel consists of four functions: "variable\_selection.R", "grad\_hess\_beta.R", "grad\_hess\_gamma.R" and "NR\_gamma.R" For further information on how to use these functions, we refer the reader to the vignette of the package.

## Details

This package consists of four functions: "variable\_selection.R", "grad\_hess\_beta.R", "grad\_hess\_gamma.R" and "NR\_gamma.R" For further information on how to use these functions, we refer the reader to the vignette of the package.

## Author(s)

Marina Gomtsyan

Maintainer: Marina Gomtsyan <mgomtsian@gmail.com>

## References

M. Gomtsyan "Variable selection in a specific regression time series of counts.", arXiv:2307.00929

## Examples

```
n = 50
p = 30
X = matrix(NA, (p+1), n)
f = 1/0.7
for(t in 1:n){X[,t] = c(1, cos(2*pi*(1:(p/2))*t*f/n), sin(2*pi*(1:(p/2))*t*f/n))}
gamma0 = c(0)
data(Y)
result = variable_selection(Y, X, gamma.init=gamma0, alpha.init=NULL, k.max=1, method="cv",
tr=0.3, n.iter=100, n.rep=1000)
beta_est = result$beta_est
Estim_active = result$estim_active
gamma_est = result$gamma_est
alpha_est = result$alpha_est
```

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grad_hess_beta	<i>Gradient and Hessian of the log-likelihood with respect to beta</i>
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**Description**

This function calculates the gradient and Hessian of the log-likelihood with respect to beta.

**Usage**

```
grad_hess_beta(Y, X, beta, gamma, alpha)
```

**Arguments**

Y	Observation matrix
X	Design matrix
beta	Initial beta vector
gamma	Initial gamma vector
alpha	Initial overdispersion parameter

**Value**

grad_L_beta	Vector of the gradient of L with respect to beta
hess_L_beta	Matrix of the Hessian of L with respect to beta

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**References**

M. Gomtsyan "Variable selection in a specific regression time series of counts.", arXiv:2307.00929

**Examples**

```
n = 50
p = 30
X = matrix(NA, (p+1), n)
f = 1/0.7
for(t in 1:n){X[,t] = c(1, cos(2*pi*(1:(p/2))*t*f/n), sin(2*pi*(1:(p/2))*t*f/n))}
gamma0 = c(0)
data(Y)
glm_nb = glm.nb(Y~t(X)[, 2:(p+1)])
beta0 = as.numeric(glm_nb$coefficients)
alpha0 = glm_nb$theta
result = grad_hess_beta(Y, X, beta0, gamma0, alpha0)
grad = result$grad_L_beta
Hessian = result$hess_L_beta
```

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grad\_hess\_gamma      *Gradient and Hessian of the log-likelihood with respect to gamma*

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### Description

This function calculates the gradient and Hessian of the log-likelihood with respect to gamma

### Usage

```
grad_hess_gamma(Y, X, beta, gamma, alpha)
```

### Arguments

Y	Observation matrix
X	Design matrix
beta	Initial beta vector
gamma	Initial gamma vector
alpha	Initial overdispersion parameter

### Value

grad_L_gamma	Vector of the gradient of L with respect to gamma
hess_L_gamma	Matrix of the Hessian of L with respect to gamma

### Author(s)

Marina Gomtsyan  
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### References

M. Gomtsyan "Variable selection in a specific regression time series of counts.", arXiv:2307.00929

### Examples

```
n = 50
p = 30
X = matrix(NA, (p+1), n)
f = 1/0.7
for(t in 1:n){X[,t] = c(1, cos(2*pi*(1:(p/2))*t*f/n), sin(2*pi*(1:(p/2))*t*f/n))}
gamma0 = c(0)
data(Y)
glm_nb = glm.nb(Y~t(X)[,2:(p+1)])
beta0 = as.numeric(glm_nb$coefficients)
alpha0 = glm_nb$theta
result = grad_hess_gamma(Y, X, beta0, gamma0, alpha0)
grad = result$grad_L_gamma
Hessian = result$hess_L_gamma
```

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NR_gamma	<i>Newton-Raphson method for estimation of gamma</i>
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**Description**

This function estimates gamma with Newton-Raphson method

**Usage**

```
NR_gamma(Y, X, beta, gamma, alpha, n.iter)
```

**Arguments**

Y	Observation matrix
X	Design matrix
beta	Initial beta vector
gamma	Initial gamma vector
alpha	Initial overdispersion parameter
n.iter	Number of iterations of the algorithm. Default=100

**Value**

gamma	Estimated gamma vector
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**Author(s)**

Marina Gomtsyan

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**References**

M. Gomtsyan "Variable selection in a specific regression time series of counts.", arXiv:2307.00929

**Examples**

```
n = 50
p = 30
X = matrix(NA, (p+1), n)
f = 1/0.7
for(t in 1:n){X[,t] = c(1, cos(2*pi*(1:(p/2))*t*f/n), sin(2*pi*(1:(p/2))*t*f/n))}
gamma0 = c(0)
data(Y)
glm_nb = glm.nb(Y~t(X)[, 2:(p+1)])
beta0 = as.numeric(glm_nb$coefficients)
alpha0 = glm_nb$theta
gamma_est = NR_gamma(Y, X, beta0, gamma0, alpha0, n.iter=100)
```

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variable\_selection      *Variable selection*

---

### Description

This function performs variable selection, estimates new vectors of beta and gamma and a new alpha

### Usage

```
variable_selection(Y, X, gamma.init, alpha.init = NULL, k.max = 1, method = "cv",
tr = 0.3, n.iter = 100, n.rep = 1000)
```

### Arguments

Y	Observation matrix
X	Design matrix
gamma.init	Initial gamma vector
alpha.init	Optional initial alpha value. The default is NULL
k.max	Number of iteration to repeat the whole algorithm
method	Stability selection method: "min" or "cv". In "min" the smallest lambda is chosen, in "cv" cross-validation lambda is chosen for stability selection. The default is "cv"
tr	Threshold for stability selection. The default is 0.3
n.iter	Number of iteration for Newton-Raphson algorithm. The default is 100
n.rep	Number of replications in stability selection step. The default is 1000

### Value

estim_active	Estimated active coefficients
beta_est	Vector of estimated beta values
gamma_est	Vector of estimated gamma values
alpha_est	Estimation of alpha

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**Examples**

```
n = 50
p = 30
X = matrix(NA, (p+1), n)
f = 1/0.7
for(t in 1:n){X[,t] = c(1, cos(2*pi*(1:(p/2))*t*f/n), sin(2*pi*(1:(p/2))*t*f/n))}
gamma0 = c(0)
data(Y)
result = variable_selection(Y, X, gamma.init=gamma0, alpha.init=NULL, k.max=1, method="cv",
tr=0.3, n.iter=100, n.rep=1000)
beta_est = result$beta_est
Estim_active = result$estim_active
gamma_est = result$gamma_est
alpha_est = result$alpha_est
```

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Y

*Observation matrix Y*

---

**Description**

An example of observation matrix

**Usage**

```
data("Y")
```

**Format**

The format is: num [1:50] 9 2 11 14 18 17 1 0 1 0 ...

**References**

M. Gomtsyan "Variable selection in a specific regression time series of counts.", arXiv:2307.00929

**Examples**

```
data(Y)
```

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