# Package 'NBtsVarSel'

July 17, 2023

Type Package
<b>Title</b> Variable Selection in a Specific Regression Time Series of Counts
Version 1.0
<b>Date</b> 2023-07-17
<b>Description</b> 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>.</arxiv:2307.00929>
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<b>Depends</b> R (>= 3.5.0), Matrix, glmnet, stats, MASS, mpath, ggplot2
VignetteBuilder knitr
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NBts VarSel-package

NBtsVarSel-package

Variable Selection in a Specific Regression Time Series of Counts

#### **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)

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#### References

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

```
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
```

grad\_hess\_beta 3

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Gradient and Hessian of the log-likelihood with respect to beta

#### **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**

Υ	Observation matrix
X	Design matrix
beta	Initial beta vector
gamma	Initial gamma vector

alpha Initial overdispertion 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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```
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
```

4 grad\_hess\_gamma

grad\_hess\_gamma

Gradient and Hessian of the log-likelihood with respect to gamma

#### **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**

Υ	Observation matrix
Χ	Design matrix
beta	Initial beta vector
gamma	Initial gamma vector
alpha	Initial overdispertion parameter

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#### 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

```
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
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Newton-Raphson method for estimation of gamma

#### **Description**

This function estimates gamma with Newton-Raphson method

#### Usage

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

#### **Arguments**

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

#### Value

gamma Estimated gamma vector

## Author(s)

Marina Gomtsyan

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#### References

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

```
\begin{array}{l} 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)\\ \end{array}
```

6 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

Υ	Observation matrix
Χ	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" $\frac{1}{2}$
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

#### Estimation of alph

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#### References

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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
```

Υ

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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