

# Package ‘GlarmaVarSel’

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**Type** Package

**Title** Variable Selection in Sparse GLARMA Models

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**Description** Performs variable selection in high-dimensional sparse GLARMA models. For further details we refer the reader to the paper Gomtsyan et al. (2020), <[doi:10.48550/arXiv.2007.08623v1](https://doi.org/10.48550/arXiv.2007.08623v1)>.

**License** GPL-2

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

**VignetteBuilder** knitr

**Suggests** knitr, markdown, formatR, doMC

**NeedsCompilation** no

**Repository** CRAN

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## Description

GlarmaVarSel 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

GlarmaVarSel 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 et al. "Variable selection in sparse GLARMA models", arXiv:2007.08623v1

## 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, gamma0, k_max=2, n_iter=100, method="min",
nb_rep_ss=1000, threshold=0.7, parallel=FALSE, nb.cores=1)
beta_est = result$beta_est
Estim_active = result$estim_active
gamma_est = result$gamma_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, beta0, gamma0)
```

**Arguments**

Y	Observation matrix
X	Design matrix
beta0	Initial beta vector
gamma0	Initial gamma vector

**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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M. Gomtsyan et al. "Variable selection in sparse GLARMA models", arXiv:2007.08623v1

**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_pois<-glm(Y~t(X)[,2:(p+1)],family = poisson)
beta0<-as.numeric(glm_pois$coefficients)
result = grad_hess_beta(Y, X, beta0, gamma0)
grad = result$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, beta0, gamma0)
```

### Arguments

Y	Observation matrix
X	Design matrix
beta0	Initial beta vector
gamma0	Initial gamma vector

### 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

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### References

M. Gomtsyan et al. "Variable selection in sparse GLARMA models", arXiv:2007.08623v1

### 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_pois<-glm(Y~t(X)[,2:(p+1)],family = poisson)
beta0<-as.numeric(glm_pois$coefficients)
result = grad_hess_gamma(Y, X, beta0, gamma0)
grad = result$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, beta0, gamma0, n_iter)
```

## Arguments

Y	Observation matrix
X	Design matrix
beta0	Initial beta vector
gamma0	Initial gamma vector
n_iter	Number of iterations of the algorithm. Default=100

## Value

gamma	Estimated gamma vector
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## Author(s)

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## References

M. Gomtsyan et al. "Variable selection in sparse GLARMA models", arXiv:2007.08623v1

## 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_pois<-glm(Y~t(X)[,2:(p+1)],family = poisson)
beta0<-as.numeric(glm_pois$coefficients)
gamma_est = NR_gamma(Y, X, beta0, gamma0, n_iter=100)
```

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variable\_selection      *Variable selection*

---

### Description

This function performs variable selection, estimates a new vector beta and a new vector gamma

### Usage

```
variable_selection(Y, X, gamma0, k_max = 2, n_iter = 100, method = "min",
  nb_rep_ss = 1000, threshold = 0.8, parallel = FALSE, nb.cores = 1)
```

### Arguments

Y	Observation matrix
X	Design matrix
gamma0	Initial gamma vector
k_max	Number of iteration to repeat the whole algorithm
n_iter	Number of iteration for Newton-Raphson algorithm
method	Stability selection method: "fast", "min" or "cv". In "min" the smallest lambda is chosen, in "cv" cross-validation lambda is chosen for stability selection. "fast" is a faster stability selection approach. The default is "min"
nb_rep_ss	Number of replications in stability selection step. The default is 1000
threshold	Threshold for stability selection. The default is 0.9
parallel	Whether to parallelize stability selection step or not. The default is FALSE
nb.cores	Number of cores for parallelization. The default is 1

### Value

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

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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, gamma0, k_max=2, n_iter=100, method="min",
nb_rep_ss=1000, threshold=0.7, parallel=FALSE, nb.cores=1)
beta_est = result$beta_est
Estim_active = result$estim_active
gamma_est = result$gamma_est
```

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Y

*Observation matrix Y*

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**Description**

An example of observation matrix

**Usage**

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

**Format**

The format is: num [1:50] 11 8 3 3 3 4 4 4 3 1 ...

**References**

M. Gomtsyan et al. "Variable selection in sparse GLARMA models", arXiv:2007.08623v1

**Examples**

```
data(Y)
```

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