

# Package ‘NGSSEML’

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**Type** Package

**Title** Non-Gaussian State-Space with Exact Marginal Likelihood

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**Description** Due to a large quantity of non-Gaussian time series and reliability data, the R-package non-Gaussian state-space with exact marginal likelihood is useful for modeling and forecasting non-Gaussian time series and reliability data via non-Gaussian state-space models with the exact marginal likelihood easily, see Gamerman, Santos and Franco (2013) <doi:10.1111/jtsa.12039> and Santos, Gamerman and Franco (2017) <doi:10.1109/TR.2017.2670142>. The package gives codes for formulating and specifying the non-Gaussian state-space models in the R language. Inferences for the parameters of the model can be made under the classical and Bayesian. Furthermore, prediction, filtering, and smoothing procedures can be used to perform inferences for the latent parameters. Applications include, e.g., count, volatility, piecewise exponential, and software reliability data.

**License** GPL (>= 2)

**URL** <https://github.com/hadht/NGSSEML-R-Package>

**Imports** mvtnorm,  
fields,  
compiler,  
dlm,  
car, interp

**Depends** R (>= 1.9.0),  
R (>= 3.5.0),  
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R (>= 3.5.0)

**Collate** 'FilteringFr'  
'gridfunction.r'  
'GridPr'  
'LikeFr'  
'LikeF2.r'

```
'ngssm.bayes.r'
'ngssm.mle.r'
'NumFail.r'
'PlotF.r'
'Prediction.r'
'PriorF.r'
'ProdXtChi.r'
'SmoothingF.r'
'TTime.r'
```

## R topics documented:

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FilteringF	<i>Filtering and One-Step-Ahead Distributions of the Latent States</i>
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---

## Description

The function FilteringF gives the shape and scale parameters of the filtering and the one-step-ahead forecast distributions of the latent states.

## Usage

```
FilteringF(formula,data,na.action="na.omit",pz=NULL,
nBreaks=NULL,model="Poisson",StaPar=NULL,a0=0.01,b0=0.01,amp=FALSE,
dist1="PRED",splot=FALSE)
```

**Arguments**

formula	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
data	a data frame containing the variables in the model. The variables are: - the time series of interest $Y_t$ (first column of the data frame). the explanatory time series to be inserted in the model. - $X_t$ must be always specified as a matrix of order $n$ by $p$ (after $Y_t$ ). - the explanatory time series to be inserted in the mean of volatility model. $Z_t$ must be always specified as a matrix of order $n$ by $p$ (after $X_t$ ). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the second column of the data frame after $Y_t$ , and the explanatory time series after the variable Event.
na.action	a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options, and is na.fail if that is unset. Optional argument.
pz	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
nBreaks	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
model	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
StaPar	a numeric vector of initial values for the static parameters. Optional argument.
a0	the shape parameter of the initial Gamma distribution. Optional argument. Default: $a_0=0.01$ .
b0	the scale parameter of the initial Gamma distribution. Optional argument. Default: $b_0=0.01$ .
amp	the interval width is taken in account in the estimation of parameter $w$ which controls the loss of information over time, only for the PEM. For more details see Santos et al. (2017). Default: FALSE. Optional argument.
distl	the latent states distribution to be returned.
splot	a plot with the point and interval estimates of the states is provided. Optional argument.

**Details**

Typical usages are

```
FilteringF(Yt~1,data=data.frame(Yt),StaPar=Par,model="Poisson",
a0=0.01,b0=0.01,splot=TRUE)
```

**Value**

att	'att' is the shape parameter of the one-step-ahead forecast distribution of the states.
btt	'btt' is the scale parameter of the one-step-ahead forecast distribution of the states.
at	'at' is the shape parameter of the filtering distribution of the states. It is necessary to specify this option in the argument 'distl'.
bt	'bt' is the scale parameter of the filtering distribution of the states. It is necessary to specify this option in the argument 'distl'.

**Note**

It is necessary to specify the argument 'distl' in order to obtain the filtering distribution of the states. The model options are the Poisson, Normal, Laplace, GED, Gamma, Weibull and Generalized Gamma models. 'Zt' are the explanatory time series only for the Normal, Laplace and GED volatility models.

**Author(s)**

T. R. Santos

**References**

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[LikeF SmoothingF](#)

**Examples**

```
library(NGSSEML)
Yt=c(1,2,1,4,3)
Par=c(0.9) #w
predpar=FilteringF(Yt~1,data=data.frame(Yt),StaPar=Par,model="Poisson",
a0=0.01,b0=0.01,splot=FALSE)
filpar=FilteringF(Yt~1,data=data.frame(Yt),StaPar=Par,model="Poisson",
a0=0.01,b0=0.01,distl="FILTER",splot=FALSE)
```

---

gridfunction

*A grid of points for obtaining the static parameters of the non-Gaussian state space models with exact marginal likelihood*

---

**Description**

The function builds a grid of the points for the static parameters of the model.

**Usage**

```
gridfunction(npoints,linf,lsup)
```

**Arguments**

npoints	the number of points/parts that the specified interval of the static parameters is partitioned.
linf	the lower limit of the static parameters in the grid.
lsup	the upper limit of the static parameters in the grid.

**Details**

Typical usages are

```
gridfunction(npoints,linf,lsup)
```

**Value**

[[1]] This function returns the grid of points for the static parameters of the model.

**Note**

The function is used to perform Bayesian inference for the static parameters. It computes the exact posterior distribution and draws a sample of the marginal posterior distribution of the static parameters using multinomial sampling scheme. It requires the R package 'fields'.

**Author(s)**

T. R. Santos

**References**

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[ngssm.bayes](#)

**Examples**

```
library(NGSSEML)
n=5 # number of points
linf=c(0,3,-1) # lower limit
lsup=c(1,6,-2) # upper limit
out=gridfunction(n,linf,lsup) # Calling the function
```

---

GridP

---

*Grid function for the PEM*


---

**Description**

The function GridP returns a vector with the interval limits (breaks) for the baseline failure rate of the PEM.

**Usage**

```
GridP(Yt, Event, nT = NULL)
```

**Arguments**

Yt	is a vector with the failure/censored times.
Event	is a censoring indicator (a vector).
nT	is the number of intervals/breaks. Optional argument. If nT==NULL, it is built one interval per failure is built.

**Details**

Typical usages are

```
GridP(Yt, Event, nT = NULL)
```

**Value**

Break                    a vector with the interval limits (breaks).

**Note**

If the argument 'nT' is 'NULL', it is built one interval per failure.

**Author(s)**

T. R. Santos

**References**

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[LikeF SmoothingF](#)

**Examples**

```
library(NGSSEML)
data(gte_data)
Yt=gte_data$V1
Xt=NULL
Zt=NULL
model="PEM"
amp=FALSE
Event=gte_data$V2
Break=GridP(Yt, Event, nT = NULL)
```

---

gte_data	<i>Daily failure times of 125 telecommunication systems installed by the GTE</i>
----------	--

---

### Description

The data are daily failure times of 125 telecommunication systems, including their respective censoring indicator, installed by the GTE corporation in a pre-specified time period (Kim and Proschan 1991).

### Usage

```
data(gte_data)
```

### Format

A data frame with 125 rows and 2 variables.

### Details

The first column of the object gte\_data corresponds to the failure times and the second to the censoring indicator.

### Source

Kim, J. S. and Proschan, R. (1991). Piecewise exponential estimator of survivor function. IEEE Transactions on Reliability, 40, 134 to 139.

### References

Kim, J. S. and Proschan, R. (1991). Piecewise exponential estimator of survivor function. IEEE Transactions on Reliability, 40, 134 to 139.

### Examples

```
data(gte_data)
```

---

LikeF	<i>Marginal Likelihood Function</i>
-------	-------------------------------------

---

### Description

This function computes the marginal likelihood function of the static parameters of the model.

### Usage

```
LikeF(formula, data, na.action="na.omit", pz=NULL,
nBreaks=NULL, model="Poisson", StaPar=NULL, a0=0.01, b0=0.01, amp=FALSE)
```

## Arguments

<code>formula</code>	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
<code>data</code>	a data frame containing the variables in the model. The variables are: - the time series of interest $Y_t$ (first column of the data frame). the explanatory time series to be inserted in the model. - $X_t$ must be always specified as a matrix of order $n$ by $p$ (after $Y_t$ ). - the explanatory time series to be inserted in the mean of volatility model. $Z_t$ must be always specified as a matrix of order $n$ by $p$ (after $X_t$ ). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the secon column of tha data frame after $Y_t$ , and he explanatory time series after the variable Event.
<code>na.action</code>	a function which indicates what should happen when the data contain NAs. The default is set by the <code>na.action</code> setting of options, and is <code>na.fail</code> if that is unset. Optional argument.
<code>pz</code>	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
<code>nBreaks</code>	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
<code>model</code>	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
<code>StaPar</code>	a numeric vector of initial values for the static parameters. Optional argument.
<code>a0</code>	the shape parameter of the initial Gamma distribution. Optional argument. Default: <code>a0=0.01</code> .
<code>b0</code>	the scale parameter of the initial Gamma distribution. Optional argument. Default: <code>b0=0.01</code> .
<code>amp</code>	the interval width is taken in account in the estimation of parameter $w$ which controls the loss of information over time, only for the PEM. For more details see Santos et al. (2017). Default: FALSE. Optional argument.

## Details

Typical usages are

```
LikeF2(Ytm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data.frame(Ytm,Xtm),model="Poisson",
StaPar=c(0.8,-0.8,0.01,0.01,0.01,0.01),a0=0.01,b0=0.01)
```

## Value

<code>llik</code>	This function returns the value of the marginal likelihood function in the logarithmic scale multiplied by -1.
-------------------	--

## Note

The model options are the Poisson, Normal, Laplace, GED, Gamma, Weibull and Generalized Gamma models. ' $Z_t$ ' are the explanatory time series only for the Normal, Laplace and GED volatility models.

## Author(s)

T. R. Santos



## References

- Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.
- Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

## See Also

[FilteringF SmoothingF](#)

## Examples

```
library(NGSSEML)
data(gte_data)
Ytm=gte_data$V1
Xtm=NULL
Ztm=NULL
Eventm=gte_data$V2
model="PEM"
LikeF(Ytm~1,data=data.frame(Ytm,Eventm),model="PEM",StaPar=c(0.8))
```

---

LikeF2

---

*Auxiliar Marginal Likelihood Function*


---

## Description

This function is an auxiliar function of the package that computes the marginal likelihood function of the transformed static parameters of the model. All transformed static parameters have values in the interval  $(-\infty, \infty)$ . The marginal likelihood function is the 'LikeF' function.

## Usage

```
LikeF2(formula, data, na.action="na.omit", pz=NULL,
nBreaks=NULL, model="Poisson", StaPar=NULL, a0=0.01, b0=0.01, amp=FALSE)
```

## Arguments

- |           |  |
|-----------|--|
| formula   | an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.   |
| data      | a data frame containing the variables in the model. The variables are: - the time series of interest Yt (first column of the data frame). the explanatory time series to be inserted in the model. - Xt must be always specified as a matrix of order n by p (after Yt). - the explanatory time series to be inserted in the mean of volatility model. Zt must be always specified as a matrix of order n by p (after Xt). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the secon column of tha data frame after Yt, and he explanatory time series after the variable Event. |
| na.action | a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options, and is na.fail if that is unset. Optional argument.   |

pz	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
nBreaks	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
model	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
StaPar	a numeric vector of initial values for the static parameters. Optional argument.
a0	the shape parameter of the initial Gamma distribution. Optional argument. Default: a0=0.01.
b0	the scale parameter of the initial Gamma distribution. Optional argument. Default: b0=0.01.
amp	the interval width is taken in account in the estimation of parameter w which controls the loss of information over time, only for the PEM. For more details see Santos et al. (2017). Default: FALSE. Optional argument.

### Details

Typical usages are

```
LikeF2(Ytm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data.frame(Ytm,Xtm),model="Poisson",
StaPar=c(log(-log(0.8)),-0.1,0.01,0.01,0.01,0.01),a0=0.01,b0=0.01)
```

### Value

llik	This function returns the value of the marginal likelihood function in the logarithmic scale multiplied by -1.
------	--

### Note

The model options are the Poisson, Normal, Laplace, GED, Gamma, Weibull and Generalized Gamma models. 'Zt' are the explanatory time series only for the Normal, Laplace and GED volatility models.

### Author(s)

T. R. Santos

### References

- Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.
- Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

### See Also

[LikeF FilteringF SmoothingF](#)

## Examples

```
library(NGSSEML)
data(gte_data)
Ytm=gte_data$V1
Xtm=NULL
Ztm=NULL
Eventm=gte_data$V2
model="PEM"
LikeF2(Ytm~1,data=data.frame(Ytm,Eventm),model="PEM",StaPar=c(log(-log(0.8))))
```

---

ngssm.bayes	<i>Bayesian estimation of the non-Gaussian state space models with exact marginal likelihood</i>
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---

## Description

The function performs the Bayesian estimation for the static parameters of the model.

## Usage

```
ngssm.bayes(formula,data,na.action="na.omit",pz=NULL,nBreaks=NULL,
model="Poisson",StaPar=NULL,amp=FALSE,a0=0.01,b0=0.01,prw=c(1,1),
prnu=NULL,prchi=NULL,prmu=NULL,prbetamu=NULL,prbetasigma=NULL,lower=NULL,
upper=NULL,ci=0.95,pointss=10,nsamplex=1000,mcmc=NULL,postplot=FALSE,contourplot=FALSE,
LabelParTheta=NULL,verbose=TRUE)
```

## Arguments

formula	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
data	a data frame containing the variables in the model. The variables are: - the time series of interest Yt (first column of the data frame). the explanatory time series to be inserted in the model. - Xt must be always specified as a matrix of order n by p (after Yt). - the explanatory time series to be inserted in the mean of volatility model. Zt must be always specified as a matrix of order n by p (after Xt). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the secon column of tha data frame after Yt, and he explanatory time series after the variable Event. The value 1 indicates failure.
na.action	a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options, and is na.fail if that is unset. Optional argument.
pz	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
nBreaks	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
model	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.

StaPar	a numeric vector of initial values for the static parameters. Optional argument.
amp	the interval width is taken in account in the estimation of parameter $w$ which controls the loss of information over time, only for the PEM. For more details see Santos et al. (2017). Default: FALSE. Optional argument.
a0	the shape parameter of the initial Gamma distribution. Optional argument. Default: a0=0.01.
b0	the scale parameter of the initial Gamma distribution. Optional argument. Default: b0=0.01.
prw	a numeric vector of length 2, indicating the hyperparameters of the Beta prior distribution for the parameter $w$ . Optional argument. The default value is c(1,1), which constitutes an uninformative prior for common data sets.
prnu	a numeric vector of length 2, indicating the hyperparameters of the Gamma prior distribution for the shape parameter $\nu$ . Optional argument.
prchi	a numeric vector of length 2, indicating the hyperparameters of the Gamma prior distribution for the shape parameter $\chi$ . Optional argument.
prmu	a numeric vector of length 2, indicating mean and standard deviation for the Gaussian prior distribution for the parameter $\mu$ . Optional argument. This prior can be used in Normal, Laplace and GED time series models.
prbetamu	a numeric vector of length $p$ , indicating mean for the Gaussian prior distribution for the parameter $\beta$ , the regression coefficients. Optional argument.
prbetasigma	a numeric matrix of order $p$ by $p$ , indicating variance-covariance matrix of the Gaussian prior distribution for the parameter $\beta$ , the regression coefficients. Optional argument.
lower	an lower bound for the static parameters (StaPar) in the density support argument of the ARMS function (MCMC). Optional argument.
upper	an upper bound for the static parameters (StaPar) in the density support argument of the ARMS function (MCMC). Optional argument.
ci	the nominal level of credibility interval for the parameters. Default: ci=0.95. Optional argument.
pointss	the number of points/parts/breaks that the specified interval of the static parameters is partitioned. Default: pointss=10.
nsamplex	the number of samples of the posterior distribution of the static parameters, obtained by numerical integration. If this posterior is computed via ARMS, nsamplex is the number of samples from the posterior distribution of the static parameters, assuming a burn-in period of 1000. Default: samples=3000.
mcmc	If true, the ARMS method is used to sample the marginal posterior distribution of the static parameters. If false, a grid of points is used to sample the marginal posterior distribution of the static parameters. Otherwise, if the mcmc argument is NULL, a suitable chose is done. Default: mcmc=NULL. Optional argument.
postplot	If true, a graph with the marginal posterior distribution of the static parameters is provided. Optional argument.
contourplot	If true, a countour plot of the posterior distribution of the static parameters is provided. Optional argument.
LabelParTheta	If not NULL, the static parameters are called by the specified label. The default value is NULL. Optional argument.
verbose	A logical variable that gives the user the output of the model fit in the console. Default: TRUE. Optional argument.

## Details

Typical usages are

```
ngssm.bayes(Ytm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data.frame(Ytm,Xtm),model=model,StaPar=c(0.8,-0.8,0.01,0.01,0.01,0.01),
prw=c(1,1),prbetamu=rep(0,5),prbetasigma=diag(10, 5, 5),pointss=5,nsamplex=1000)
```

## Value

[[1]]	This function returns the output of Bayesian estimation for the static parameters.
[[2]]	This function returns posterior samples of the static parameters using multinomial sampling scheme.

## Note

This function provides summaries of the posterior distribution of the static parameters of the specified model. In an exact way, the posterior is built to make inferences for the static parameters, and samples of it are drawn using multinomial sampling. If the dimensionality of static parameters and the break number of the grid are high, there are many points to evaluate the posterior distribution and, hence, an MCMC method (ARMS) is used to sample the posterior distribution of the static parameters. Furthermore, it is necessary to specify the limits of the parametric space of the model for the ARMS function in the arguments 'lower' and 'upper'.

## Author(s)

T. R. Santos

## References

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

## See Also

[PriorF SmoothingF ngssm.mle](#)

## Examples

```
#####
##
## PEM Example: the GTE data
##
#####
library(NGSSEML)
#### Inputs:
data(gte_data)
Ytm=gte_data$V1
Eventm=gte_data$V2 # Event: failure, 1.
Breakm=GridP(Ytm, Eventm, nT = NULL)
Xtm=NULL
Ztm=NULL
```

```

model="PEM"
amp=FALSE
#LabelParTheta=c("w")
StaPar=c(0.5)
p=length(StaPar)
nn=length(Ytm)
a0=0.01
b0=0.01
#pointss=500000    ### points
pointss=10    ### points
nsamplex=1000 ## Sampling posterior
ci=0.95
alpha=1-ci
#Fit:
fitbayes=ngssm.bayes(Ytm~1,data=data.frame(Ytm,Eventm),model=model,pz=NULL,
StaPar=StaPar,amp=amp,a0=a0,b0=b0,prw=c(1,1),prnu=NULL,prchi=NULL,prmu=NULL,
prbetamu=NULL,prbetasigma=NULL,ci=ci,pointss=pointss,nsamplex=nsamplex,
postplot=FALSE,contourplot=FALSE)

#####
##
##  SR WEIBULL MODEL: the SYS1 data
##
#####
#library(NGSSEML)
#### Defaults values (NULL):
#### Inputs:
data(sys1_data)
Ytm=sys1_data[,1]+0.00001
Xtm=sys1_data[,2]
model="SRWeibull"
#LabelParTheta=c("w","nu","Beta")
pointss=5    ### points
##Fit:
StaPar=c(0.98,0.75,0.02)
fitbayes=ngssm.bayes(Ytm~Xtm,data=data.frame(Ytm,Xtm),
model=model,pz=NULL,StaPar=StaPar,
prw=c(1,1),prnu=c(0.1,0.1),prbetamu=rep(0,1),prbetasigma=diag(100,1,1),
pointss=pointss,nsamplex=3000,postplot=FALSE,contourplot=FALSE)
#####
#####

```

---

ngssm.mle

---

*Maximum likelihood estimation of the non-Gaussian state space models with exact marginal likelihood*


---

## Description

The function performs the marginal likelihood estimation for the static parameters of the model.

## Usage

```

ngssm.mle(formula, data,na.action="na.omit",pz=NULL,
nBreaks=NULL,model="Poisson",StaPar=NULL,amp=FALSE,
a0=0.01,b0=0.01,ci=0.95,LabelParTheta=NULL,verbose=TRUE)

```

## Arguments

<code>formula</code>	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
<code>data</code>	a data frame containing the variables in the model. The variables are: - the time series of interest $Y_t$ (first column of the data frame). the explanatory time series to be inserted in the model. - $X_t$ must be always specified as a matrix of order $n$ by $p$ (after $Y_t$ ). - the explanatory time series to be inserted in the mean of volatility model. $Z_t$ must be always specified as a matrix of order $n$ by $p$ (after $X_t$ ). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the second column of the data frame after $Y_t$ , and the explanatory time series after the variable Event. The value 1 indicates failure.
<code>na.action</code>	a function which indicates what should happen when the data contain NAs. The default is set by the <code>na.action</code> setting of options, and is <code>na.fail</code> if that is unset. Optional argument.
<code>pz</code>	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
<code>nBreaks</code>	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
<code>model</code>	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
<code>StaPar</code>	a numeric vector of initial values for the static parameters. Optional argument.
<code>amp</code>	the interval width is taken in account in the estimation of parameter $w$ which controls the loss of information over time, only for the PEM. For more details see Santos et al. (2017). Default: FALSE. Optional argument.
<code>a0</code>	the shape parameter of the initial Gamma distribution. Optional argument. Default: $a_0=0.01$ .
<code>b0</code>	the scale parameter of the initial Gamma distribution. Optional argument. Default: $b_0=0.01$ .
<code>ci</code>	the nominal level of confidence interval for the parameters. Default: $ci=0.95$ . Optional argument.
<code>LabelParTheta</code>	If not NULL, the static parameters are called by the specified label. Optional argument.
<code>verbose</code>	A logical variable that gives the user the output of the model fit in the console. Default: TRUE. Optional argument.

## Details

Typical usages are

```
fit=ngssm.mle(Ytm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data1,model="Poisson",StaPar=c(0.8,-0.8,0.01,0.01,0.01,0.01),
a0=0.01,b0=0.01,ci=0.95)
```

## Value

<code>[[1]]</code>	the output of the model fit, presenting the maximum likelihood estimators, standard errors, Z statistics, and asymptotic confidence intervals of the model parameters.
--------------------	--

**Note**

The function provides the MLE estimates for the static parameters of the specified model. The likelihood function is maximized using the 'optim' function and 'BFGS' method.

**Author(s)**

T. R. Santos

**References**

- Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.
- Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[FilteringF SmoothingF ngssm.bayes](#)

**Examples**

```
#####
#
## PEM Example: the GTE data
##
#####
# MLE estimation:
library(NGSSEML)
data(gte_data)
Ytm=gte_data$V1
Xtm=NULL
Ztm=NULL
model="PEM"
amp=FALSE
Eventm=gte_data$V2      # Event: failure, 1.
Break=GridP(Ytm, Eventm, nT = NULL)
#LabelParTheta=c("w")
StaPar=c(0.73)
a0=0.01
b0=0.01
ci=0.95
fit=ngssm.mle(Ytm~1,data=data.frame(Ytm,Eventm),model=model,nBreaks=NULL,
amp=amp,a0=a0,b0=b0,ci=ci)

#####
##
## SR GAMMA MODEL: the SYS1 data
##
#####
# MLE estimation:
#library(NGSSEML)
data(sys1_data)
Ytm=sys1_data[,1]+0.00001
Xtm=sys1_data[,2]      # Xt as matrix always!
Zt=NULL
```



```

model="SRWeibull"
#LabelParTheta=c("w","alpha","Beta1")
StaPar=c(0.9,0.7,0.01)
fit=ngssm.mle(Ytm~Xtm,data=data.frame(Ytm,Xtm),
model=model,pz=NULL,StaPar=StaPar,a0=0.01,b0=0.01,ci=0.95)

```

---

NumFail	<i>Number of failures in each interval</i>
---------	--

---

## Description

The function NumFail gives the number of failures in each interval of the PEM.

## Usage

```
NumFail(StaPar,Yt,Event,Break,Xt)
```

## Arguments

StaPar	is the static parameter vector.
Yt	is a vector with the failure/censored times.
Event	is a censoring indicator (a vector).
Break	is a vector with the interval limits.
Xt	are the explanatory variables to be inserted in the model.

## Details

Typical usages are

```
NumFail(StaPar,Yt,Event,Break,Xt)
```

## Value

nf 'nf' is a vector with the number of failures in each interval.

## Note

The function provides the number of failures in each interval.

## Author(s)

T. R. Santos

## References

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[LikeF SmoothingF](#)

**Examples**

```
#####
#
## PEM Example: the GTE data
##
#####
library(NGSSEML)
data(gte_data)
Yt=gte_data$V1
Xt=NULL
Zt=NULL
amp=FALSE
Event=gte_data$V2      # Event: failure, 1.
Break=GridP(Yt, Event, nT = NULL)
StaPar=c(0.7)
NumFail(StaPar,Yt,Event, Break, Xt = NULL)
```

---

PlotF

---

*Plot Function*


---

**Description**

The function PlotF gives graphs with smoothed/filtered estimates of the latent states.

**Usage**

```
PlotF(formula, data, na.action="na.omit", pz=NULL, nBreaks=NULL,
plotYt=TRUE, axisxdate=NULL, transf=1, model="Poisson", posts, Proc="Smooth",
Type="Marg", distl="PRED", a0=0.01, b0=0.01, ci=0.95, startdate=NULL, enddate=NULL,
Freq=NULL, Typeline='l', cols=c("black", "blue", "lightgrey"), xlab="t",
yylab=expression(paste(hat(mu)[t])), xlim=NULL, ylim=NULL, Lty=c(1,2,1),
Lwd=c(2,2,2), Cex=0.68)
```

**Arguments**

formula	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
data	a data frame containing the variables in the model. The variables are: - the time series of interest Yt (first column of the data frame). the explanatory time series to be inserted in the model. - Xt must be always specified as a matrix of order n by p (after Yt). - the explanatory time series to be inserted in the mean of volatility model. Zt must be always specified as a matrix of order n by p (after Xt). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the secon column of tha data frame after Yt, and he explanatory time series after the variable Event. The value 1 indicates failure.

na.action	a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options, and is na.fail if that is unset. Optional argument.
pz	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
nBreaks	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
transf	This argument allows the user to apply a transformation (exponentiation) in the estimates of the latent states. For example, the inverse transformation, i. e., $\text{transf} = -1$ . The default value is 1. Optional argument.
model	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
posts	A sample or an estimate of the static parameters.
plotYt	If true, the time series $Y_t$ is inserted in the plot. The default value is TRUE. Optional argument.
axisxdate	a date vector for the x-axis can be specified in this function. The default value is NULL. Optional argument.
Proc	the chosen distribution of the latent states. There are 2 options: conditional ("Cond") on the static parameters and marginal ("Marg"). The default is conditional ("Marg").
Type	the latent states distribution to be returned. There are 2 options: the smoothed ("Smooth") and filtering ("Filter") distributions. the chosen distribution of the latent states. There are 2 options: conditional ("Cond") on the static parameters and marginal ("Marg"). The default is conditional ("Marg").
dist1	the chosen distribution of the latent states. There are 2 options: conditional on the static parameters and marginal ("Marg"). The default is conditional ("Cond").
a0	the shape parameter of the initial Gamma distribution. Optional argument. Default: $a_0=0.01$ .
b0	the scale parameter of the initial Gamma distribution. Optional argument. Default: $b_0=0.01$ .
ci	the nominal level of confidence interval for the parameters. Optional argument. Default: $ci=0.95$ .
startdate	If the argument axisxdate is not NULL, it is necessary to specify a start date. Optional argument.
enddate	If the argument axisxdate is not NULL, it is necessary to specify an end date. Optional argument.
Freq	If the argument axisxdate is not NULL, it is necessary to specify a frequency of the data. Optional argument.
Typeline	the type of plot should be drawn. Possible types are "p" for points, "l" for lines, "s" for stair steps and etc. Optional argument.
cols	You can specify colors in the graph. Optional argument.
xxlab	a title for the x-axis. Optional argument.
yylab	a title for the y-axis. Optional argument.
xxlim	a numeric vector with limits for the x-axis. Optional argument.

yylim	a numeric vector with limits for the y-axis. Optional argument.
Lty	A line type. Optional argument.
Lwd	Line width relative to the default (default=1). 2 is twice as wide. Optional argument.
Cex	number indicating the amount by which plotting text and symbols should be scaled relative to the default. 1=default, 1.5 is 50% larger, 0.5 is 50% smaller, etc. . Optional argument.

## Details

Typical usages are

```
PlotF(YYtm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data.frame(Ytm,Xtm),model="Poisson",StaPar=estopt,axisxdate=x,Proc="Smooth",
Type="Cond",distl="FILTER",a0=0.01,b0=0.01,ci=0.95,posts=estopt,
startdate="1970/01/01",enddate="1983/12/31",Freq="months",
cols=c("black","blue","lightgrey"),xxlab="t",yylabel="Yt",yylim=c(0,15),
Lty=c(1,2,1),Lwd=c(2,2,2),Cex=0.68
```

## Value

graph	This function returns an graph with smoothed or filtered estimates of the latent states.
-------	--

## Note

The model options are the Poisson, Normal, Laplace, GED, Gamma, Weibull and Generalized Gamma models. 'Zt' are the explanatory time series only for the Normal, Laplace and GED volatility models.

## Author(s)

T. R. Santos

## References

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

## See Also

[FilteringF](#) [SmoothingF](#) [ngssm.bayes](#) [ngssm.mle](#)

## Examples

```
#####
## Petro data:
library(NGSSEML)
#### Inputs:
data(Rt)
```

```

Ytm=Rt$Rt
Date=Rt$Date
Xtm=NULL
Ztm=NULL
model="GED"
LabelParTheta=c("W", "nu")
StaPar=c(0.9,1)
p=length(StaPar)
nn=length(Ytm)
a0=0.01
b0=0.01
pointss=5    ### points
nsamplex=50 ## Sampling posterior
ci=0.95      # Cred. level
fitbayes=ngssm.bayes(Ytm~1,data=data.frame(Ytm),model=model,pz=NULL,
StaPar=StaPar,a0=a0,b0=b0,prw=c(1,1),
prnu=c(0.01,0.01),ci=ci,pointss=pointss,nsamplex=nsamplex,
postplot=FALSE,contourplot=FALSE,LabelParTheta=LabelParTheta)
posts=fitbayes[[2]]
#####
#Smoothing:
#####
#PlotF function:
PlotF(Ytm~1,data=data.frame(Ytm),model=model,pz=NULL,plotYt=FALSE,
transf=-0.5,Proc="Smooth",Type="Marg",distl="PRED",a0=a0,b0=b0,
ci=ci,posts=posts,startdate=NULL,enddate=NULL,Freq="days",Typeline='l',
cols=c("black","blue","lightgrey"),xxlab="t",yylab=expression(paste(hat(sigma)[t])),
yylim=c(0.02,0.10),Lty=c(1,2,1),Lwd=c(2,2,2),Cex=0.68)
# dev.new()
#####

```

---

Prediction

---

*The h-step-ahead forecast for the observations*


---

## Description

The function Prediction implements the h-step-ahead forecast for the observations of SR and Poisson models.

## Usage

```

Prediction(formula,data,na.action="na.omit",pz=NULL,nBreaks=NULL,
model="Poisson",StaPar=NULL,a0=0.01,b0=0.01,distl="PRED",
ci=0.95,samples=500,hh=1,Xtprev=NULL,method="MLE")

```

## Arguments

formula	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
data	a data frame containing the variables in the model. The variables are: - the time series of interest Yt (first column of the data frame). the explanatory time series to be inserted in the model. - Xt must be always specified as a matrix of order n by p (after Yt). - the explanatory time series to be inserted in the mean of

	volatility model. $Z_t$ must be always specified as a matrix of order $n$ by $p$ (after $X_t$ ). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the second column of the data frame after $Y_t$ , and the explanatory time series after the variable Event. The value 1 indicates failure.
na.action	a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options, and is na.fail if that is unset. Optional argument.
pz	the number of the explanatory time series to be inserted in the mean of volatility model. Default: NULL. Optional argument.
nBreaks	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
model	the chosen model for the observations. The options are: Poisson, SRGamma, and SRWeibull models.
StaPar	a numeric vector of initial values for the static parameters. Optional argument.
a0	the shape parameter of the initial Gamma distribution. Optional argument. Default: $a_0=0.01$ .
b0	the scale parameter of the initial Gamma distribution. Optional argument. Default: $b_0=0.01$ .
dist1	the latent states distribution to be returned. Optional argument.
ci	the confidence level of the confidence interval for the states. Optional argument. Default: $ci=0.95$ .
samples	the number of samples drawn from the predictive distributions for the observations, given a point of the static parameters (StaPar). Default: $samples=500$ .
hh	the forecast horizon. Optional argument. Default: $hh=1$ .
Xtprev	the future values of covariates, if there are covariates in the model. Optional argument.
method	the estimation method used, MLE or Bayesian.

## Details

Typical usages are

```
Prediction(Ytm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data1,model=model,pz=NULL,StaPar=estopt,a0=a0,b0=b0,ci=ci,dist1="PRED",
samples=500,hh=hh,Xtprev=Xtprev,method="MLE")
```

## Value

Mean	'Mean' is the mean of the one-step-ahead forecast distribution for the observations.
Median	'Median' is the median of the one-step-ahead forecast distribution for the observations.
Perc1	'Perc1' the the percentile of $((1-ci)/2)$ order of the one-step-ahead forecast distribution for the observations.
Perc2	'Perc2' is the percentile of $(1-(1-ci)/2)$ order of the one-step-ahead forecast distribution for the observations.

**Note**

The model options are Poisson, SRWeibull and SRGamma models, because the remaining are the volatility and piecewise exponential/PH models.

**Author(s)**

Thiago Rezende dos Santos

**References**

- Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.
- Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[LikeF SmoothingF](#)

**Examples**

```
#####
##
##  SR Weibull MODEL: the SYS1 data
##
#####
library(NGSSEML)
#Classical:
data(sys1_data)
Ytm=sys1_data[,1]+0.00001
Xtm=as.matrix(sys1_data[,2])
hh=3
Xtprev=as.matrix(135+1:hh)
model="SRWeibull"
#model="SRGamma"
LabelParTheta=c("w", "nu", "Beta1")
StaPar=c(0.9,0.7,0.01)
#Fit:
fit=ngssm.mle(Ytm~Xtm,data=data.frame(Ytm,Xtm),
model=model,pz=NULL,StaPar=StaPar,a0=0.01,b0=0.01,ci=0.95)
estopt=c(0.99,0.75262104,0.02342691)
estopt=fit[[1]][1:3]
predpar=Prediction(Ytm~Xtm,data=data.frame(Ytm,Xtm),
model=model,StaPar=estopt,pz=NULL,a0=0.01,
b0=0.01,dist1="FILTER",ci=0.95,samples=500,hh=hh,
Xtprev=Xtprev,method="MLE")

#library(NGSSEML)
#Bayesian:
#### Inputs:
data(sys1_data)
Yt=sys1_data[,1]+0.00001
Xt=as.matrix(sys1_data[,2]) # Xt as matrix always!
hh=3
```

```

Xtprev=as.matrix(135+1:hh)
Zt="NULL"
model="SRWeibull"
#model="SRGamma"
LabelParTheta=c("w","nu","Beta1")
pointss=4    ### points
nsamplex=50 ## Multinomial sampling posterior
StaPar=c(0.9,0.7,0.01)
#Fit:
fitbayes=ngssm.bayes(Ytm~Xtm,data=data.frame(Ytm,Xtm),
model=model,pz=NULL,StaPar=StaPar,
prw=c(1,1),prnu=c(0.1,0.1),prbetamu=rep(0,1),prbetasigma=diag(100,1,1),
pointss=pointss,nsamplex=50,postplot=FALSE,contourplot=FALSE)
posts=fitbayes[[2]]
#Prediction:
set.seed(1000)
predpar=Prediction(Ytm~Xtm,data=data.frame(Ytm,Xtm),
model=model,pz=NULL,StaPar=posts,
a0=0.01,b0=0.01,dist1="PRED",ci=0.95,samples=500,hh=hh,
Xtprev=Xtprev,method="Bayes")
predpar

```

PriorF

*Prior Function*

## Description

This function computes the probability density function of the specified prior for the static parameters of the model.

## Usage

```
PriorF(StaPar,model="Poisson",prw=c(1,1),prnu=NULL,prchi=NULL,prmu=NULL,
prbetamu=NULL,prbetasigma=NULL)
```

## Arguments

StaPar	the static parameter vector.
model	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
prw	the numeric vector of length 2, indicating the hyperparameters of the Beta prior distribution for the parameter w. Optional argument. The default value is c(1,1), which constitutes an uninformative prior for common data sets.
prnu	the numeric vector of length 2, indicating the hyperparameters of the Gamma prior distribution for the shape parameter nu. Optional argument. The default value is NULL.
prchi	the numeric vector of length 2, indicating the hyperparameters of the Gamma prior distribution for the shape parameter chi. Optional argument. The default value is NULL.



prmu	the numeric vector of length 2, indicating the mean and standard deviation for the Gaussian prior distribution for the parameter mu. Optional argument. The default value is NULL. This prior can be used in Normal, Laplace and GED time series models.
prbetamu	the numeric vector of length p, indicating mean for the Gaussian prior distribution for the parameter beta, the regression coefficients. Optional argument. The default value is NULL.
prbetasigma	the numeric matrix of order p by p, indicating variance-covariance matrix of the Gaussian prior distribution for the parameter beta, the regression coefficients. Optional argument. The default value is NULL.

### Details

Typical usages are

```
PriorF(StaPar,model="Poisson",prw=c(1,1),prnu=NULL,prchi=NULL,
prmu=NULL,prbetamu=NULL,prbetasigma=NULL)
```

### Value

llik This function returns the probability density of the specified prior for the static parameters of the model.

### Note

The model options are the Poisson, Normal, Laplace, GED, Gamma, Weibull and Generalized Gamma models. The arguments of this function depend on the static parameters in each model to be specified to the data.

### Author(s)

T. R. Santos

### References

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

### See Also

[ngssm.bayes SmoothingF](#)

### Examples

```
library(NGSSEML)
#Trend,CosAnnual,SinAnnual,CosSemiAnnual,SinSemiAnnual
LabelParTheta=c("w","Beta1","Beta2","Beta3","Beta4","Beta5")
StaPar=c(0.8,-0.1,0.01,0.01,0.01,0.01)
PriorF(StaPar,model="Poisson",prw=c(1,1),prnu=NULL,prchi=NULL,prmu=NULL,
prbetamu=rep(0,5),prbetasigma=diag(4, 5, 5))
```

---

ProdXtChi

*ProdXtChi*


---

**Description**

ProdXtChi computes a constant to be inserted in the likelihood of the PEM with covariate (PH model).

**Usage**

```
ProdXtChi(StaPar,Yt,Break,Event,Xt)
```

**Arguments**

StaPar	the static parameter vector.
Yt	a vector with the failure/censored times.
Break	a vector with the interval limits.
Event	a censoring indicator (a vector).
Xt	the explanatory variables to be inserted in the model.

**Details**

Typical usages are

```
ProdXtChi(StaPar,Yt,Break,Event,Xt)
```

**Value**

cco	a constant.
-----	-------------

**Note**

This function is ONLY to compute a constant to be inserted in the likelihood of the PEM with covariate. It is more an internal function.

**Author(s)**

T. R. Santos

**References**

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

**See Also**

[LikeF SmoothingF](#)

## Examples

```
library(NGSSEML)
data(gte_data)
Yt=gte_data$V1
Xt=as.matrix(1:125)
model="PEM"
amp=FALSE
Event=gte_data$V2      # Event: failure, 1.
Break=GridP(Yt, Event, nT = NULL)
LabelParTheta=c("w")
StaPar=c(0.73)
out=ProdXtChi(StaPar, Yt, Break, Event, Xt)
```

---

Rt	<i>Returns of the asset PETR3 (Petrobras company) in the Brazilian stock market</i>
----	---

---

## Description

The return data consist of 1999 daily observations in the period of 2000/01/06 to 2008/29/01.

## Usage

```
data(Rt)
```

## Format

A data frame with 1999 rows and 1 variable.

## Details

The data irregularity due to weekends and holidays was ignored.

## Source

<http://http://finance.yahoo.com/>

## References

<https://br.advfn.com/bolsa-de-valores/bovespa/petrobras-PETR3/empresa>

## Examples

```
data(Rt)
```

## Description

The function `SmoothingF` gives an exact sample of the posterior distribution of the latent states conditional on the static parameters or marginal.

## Usage

```
SmoothingF(formula,data,na.action="na.omit",pz=NULL,nBreaks=NULL,
model="Poisson",StaPar=NULL,Type="Cond",a0=0.01,b0=0.01,
amp=FALSE,samples=1,ci=0.95,plot=FALSE)
```

## Arguments

<code>formula</code>	an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
<code>data</code>	a data frame containing the variables in the model. The variables are: - the time series of interest $Y_t$ (first column of the data frame). the explanatory time series to be inserted in the model. - $X_t$ must be always specified as a matrix of order $n$ by $p$ (after $Y_t$ ). - the explanatory time series to be inserted in the mean of volatility model. $Z_t$ must be always specified as a matrix of order $n$ by $p$ (after $X_t$ ). - a censoring indicator of the event (a vector), only for the PEM. If the model is the PEM, put the variable Event in the second column of the data frame after $Y_t$ , and the explanatory time series after the variable Event. The value 1 indicates failure.
<code>na.action</code>	a function which indicates what should happen when the data contain NAs. The default is set by the <code>na.action</code> setting of options, and is <code>na.fail</code> if that is unset. Optional argument.
<code>pz</code>	the number of the explanatory time series to be inserted in the mean of volatility model. Default: <code>NULL</code> . Optional argument.
<code>nBreaks</code>	the number of breaks used to build a vector with the interval limits, only for the PEM. Optional argument.
<code>model</code>	the chosen model for the observations. The options are: Poisson, Normal, Gamma, Weibull, Generalized Gamma, Laplace, GED and PEM models.
<code>StaPar</code>	a numeric vector of initial values for the static parameters. Optional argument.
<code>Type</code>	the chosen distribution of the latent states. There are 2 options: conditional on the static parameters and marginal ("Marg"). The default is conditional ("Cond").
<code>a0</code>	the shape parameter of the initial Gamma distribution. Optional argument. Default: <code>a0=0.01</code> .
<code>b0</code>	the scale parameter of the initial Gamma distribution. Optional argument. Default: <code>b0=0.01</code> .
<code>amp</code>	the interval width is taken in account in the estimation of parameter $w$ which controls the loss of information over time, only for the PEM. For more details see Santos et al. (2017). Default: <code>FALSE</code> . Optional argument.

<code>samples</code>	the number of samples drawn from the joint posterior distribution of the latent states, given a point of the static parameters (StaPar). Optional argument. Default: <code>samples = 1</code> .
<code>ci</code>	the nominal level of confidence interval for the parameters. Optional argument. Default: <code>ci=0.95</code> .
<code>splot</code>	Create a plot with the point and interval estimates of the states. Optional argument.

## Details

Typical usages are

```
SmoothingF(Ytm~Trend+CosAnnual+SinAnnual+CosSemiAnnual+SinSemiAnnual,
data=data.frame(Ytm,Xtm),model="Poisson",Type="Cond",a0=0.01,b0=0.01,samples=1,ci=0.95)
```

## Value

`mdata` This function returns an exact sample of the join distribution of the states. If the number of samples is greater than 1, some summaries of the state samples are returned.

## Note

The model options are the Poisson, Normal, Laplace, GED, Gamma, Weibull and Generalized Gamma models. 'Zt' are the explanatory time series only for the Normal, Laplace and GED volatility models.

## Author(s)

T. R. Santos

## References

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

## See Also

[FilteringF LikeF ngssm.mle ngssm.bayes](#)

## Examples

```
#####
##
## CLASSICAL ESTIMATION
##
#####

#####
##
## PEM Example: the GTE data
```

```

##
#####
library(NGSSEML)
data(gte_data)
Ytm=gte_data$V1
Xtm=NULL
Ztm=NULL
model="PEM"
amp=FALSE
Eventm=gte_data$V2
Breakm=GridP(Ytm, Eventm, nT = NULL)
LabelParTheta=c("w")
StaPar=c(0.73)
a0=0.01
b0=0.01
ci=0.95
fit=ngssm.mle(Ytm~1,data=data.frame(Ytm,Eventm),model=model,StaPar=StaPar,
nBreaks=NULL,amp=amp,a0=a0,b0=b0,ci=ci,LabelParTheta=LabelParTheta)
estopt=fit[[1]][1]
set.seed(10)
MeanSmooth=SmoothingF(Ytm~1,data=data.frame(Ytm,Eventm),model=model,
a0=0.01,b0=0.1,amp=FALSE,samples=1000,ci=0.95,splo=FALSE,StaPar=estopt)

#Smoothing:
set.seed(1000)
fits=SmoothingF(Ytm~1,data=data.frame(Ytm,Eventm),model=model,Type="Cond",
a0=0.01,b0=0.1,ci=0.95,samples=1,splo=FALSE,StaPar=estopt)
##PlotF function:
PlotF(Ytm~1,data=data.frame(Ytm,Eventm),model=model,plotYt=FALSE,
axisxdate=Breakm[1:17],Proc="Smooth",Type="Cond",distl="Filter",a0=0.01,b0=0.1,
ci=0.95,posts=estopt,Typeline='s',
cols=c("black","blue","lightgrey"),xxlab="Time to Failure (Days)",yylab="Failure rate",
yylim=c(0,0.008),xxlim=c(0,139),Lty=c(1,2,1),Lwd=c(2,2,2),Cex=0.68)

#####
##
## BAYESIAN ESTIMATION
##
#####

#####
##PEM
##GTE Data
#####
#library(NGSSEML)
### Inputs:
data(gte_data)
Ytm=gte_data$V1
Eventm=gte_data$V2
Breakm=GridP(Ytm, Eventm, nT = NULL)
Xtm=NULL
Ztm=NULL
model="PEM"
amp=FALSE
LabelParTheta=c("w")
StaPar=c(0.73)
p=length(StaPar)

```

```

nn=length(Breakm)
a0=0.01
b0=0.1
p=length(StaPar)
pointss=5    ### points
nsamplex=100 ## Multinomial sampling posterior
ci=0.95
alpha=1-ci
#Fit:
#Bayesian:
fitbayes=ngssm.bayes(Ytm~1,data=data.frame(Ytm,Eventm),model=model,
pz=NULL,StaPar=StaPar,amp=amp,a0=a0,b0=b0,prw=c(1,1),prnu=NULL,prchi=NULL,
prmu=NULL,prbetamu=NULL,prbetasigma=NULL,ci=ci,pointss=pointss,nsamplex=nsamplex,
postplot=FALSE,contourplot=FALSE,LabelParTheta=LabelParTheta)
posts=fitbayes[[2]]
#Smoothing:
set.seed(1000)
fits=SmoothingF(Ytm~1,data=data.frame(Ytm,Eventm),model=model,pz=NULL,
StaPar=posts,Type="Marg",a0=a0,b0=b0,ci=ci,samples=1,splo=FALSE)
#####

```

---

sys1\_data

---

*The times between successive computer software failures of the SYS1*


---

## Description

The times between 136 successive computer software failures and the number of failures of the SYS1 data.

## Usage

```
data(sys1_data)
```

## Format

A data frame with 136 rows and 2 variables.

## Details

The first column of the object sys1\_data corresponds to the times and the second to the number of detected failures before the i-th stage.

## Source

Lyu, M. R. (1996). Handbook of software reliability engineering.

## References

Lyu, M. R. (1996). Handbook of software reliability engineering.

## Examples

```
data(sys1_data)
```

---

TTime	<i>Total Failure Time</i>
-------	---------------------------

---

### Description

The function TTime gives a vector with the contribution of the units for each interval of the PEM.

### Usage

```
TTime(StaPar,Yt,Event, Break, Xt = NULL)
```

### Arguments

StaPar	the static parameter vector.
Yt	a vector with the failure/censored times.
Event	a censoring indicator (a vector).
Break	a vector with the interval limits.
Xt	the explanatory variables to be inserted in the model. Optional argument.

### Details

Typical usages are

```
TTime(StaPar,Yt,Event, Break, Xt = NULL)
```

### Value

ttime	a vector with the contribution of the units for each interval.
-------	--

### Note

This function provides the total failure time.

### Author(s)

T. R. Santos

### References

Gamerman, D., Santos, T. R., and Franco, G. C. (2013). A Non-Gaussian Family of State-Space Models with Exact Marginal Likelihood. *Journal of Time Series Analysis*, 34(6), 625-645.

Santos T. R., Gamerman, D., Franco, G. C. (2017). Reliability Analysis via Non-Gaussian State-Space Models. *IEEE Transactions on Reliability*, 66, 309-318.

### See Also

[LikeF SmoothingF](#)



### Examples

```
#####
#
## PEM Example: the GTE data
##
#####
library(NGSSEML)
data(gte_data)
Yt=gte_data$V1
Xt=NULL
Zt=NULL
amp=FALSE
Event=gte_data$V2      # Event: failure, 1.
Break=GridP(Yt, Event, nT = NULL)
StaPar=c(0.7)
out=TTime(StaPar,Yt,Event, Break, Xt = NULL)
```

Yt

*The Polio Data*

### Description

The data consist of monthly counts of poliomyelitis cases in the USA from the year 1970 to 1983.

### Usage

```
data(Yt)
```

### Format

A data frame with 168 observations on the following 8 variables.

### Details

The covariates are the deterministic trend centered at 73 and divided by 1000, annual and semiannual cosine and annual and semiannual sine.

### Source

Centers for Disease Control, USA.

### References

Zeger, S.L. (1988). A regression model for time series of counts. *Biometrika* 75, 621-29.

### Examples

```
data(Yt)
```

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