# Package 'saemix'

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Type Package
Title Stochastic Approximation Expectation Maximization (SAEM) algorithm
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Description The SAEMIX package implements the Stochastic Approximation EM algorithm for parameter estimation in (non)linear mixed effects models. The SAEM algorithm: - computes the maximum likelihood estimator of the population parameters, without any approximation of the model (linearisation, quadrature approximation,), using the Stochastic Approximation Expectation Maximization (SAEM) algorithm, - provides standard errors for the maximum likelihood estimator - estimates the conditional modes, the conditional means and the conditional standard deviations of the individual parameters, using the Hastings-Metropolis algorithm. Several applications of SAEM in agronomy, animal breeding and PKPD analysis have been published by members of the Monolix group (http://group.monolix.org/).
License GPL (>= 2)
LazyLoad yes
LazyData yes
Imports graphics, stats, methods
Collate global.R SaemixData.R SaemixModel.R SaemixRes.R SaemixObject.R main.R zzz.R
R topics documented:
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# **Description**

- Computing the maximum likelihood estimator of the population parameters, without any approximation of the model (linearization, quadrature approximation, . . . ), using the Stochastic Approximation Expectation Maximization (SAEM) algorithm

- Estimation of the Fisher Information matrix
- Estimation of the individual parameters
- Estimation of the likelihood
- Plot convergence graphs

#### **Details**

Package: saemix
Type: Package
Version: 0.9

Date: 2010-09-19 License: GPL (>=) 1.2

LazyLoad: yes

The SAEM package includes a number of undocumented functions, which are not meant to be used directly by the user.

default setdefault

**computational functions** cutoff,cutoff.max, cutoff.eps, cutoff.res, compute.Uy, compute.Uy.nocov, conditional.distribution, gqg.mlx

distributions normcdf, norminv

error model error

sampling trnd.mlx, tpdf.mlx, gammarnd.mlx

parameter transformations transpsi, transphi, dtransphi

## Author(s)

Emmanuelle Comets <a href="mailto:comets@inserm.fr">emmanuelle.comets@inserm.fr</a>>, Audrey Lavenu, Marc Lavielle.

# References

Kuhn, E., and Lavielle, M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038. Monolix32\_UsersGuide.pdf (http://software.monolix.org/sdoms/software/)

#### See Also

nlme, Saemix Data, Saemix Model, Saemix Object, saemix

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## **Examples**

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,dimnames=list(NULL, byrow=1,0.00)
 c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE), \ error.model="constant")
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# print(saemix.fit)
# plot(saemix.fit)
```

coef-methods

Methods for Function coef

## **Description**

Methods for function coef

#### Methods

```
signature(x = "ANY") default coef function ?
signature(x = "SaemixObject") extracts coefficients from an SaemixObject
```

conddist.saemix 5

conddist.saemix	Estimate conditional mean and variance of individual parameters using the MCMC algorithm

# **Description**

When the parameters of the model have been estimated, we can estimate the individual parameters (psi\_i).

Let hattheta be the estimated value of theta computed with the SAEM algorithm and let p(phi\_i ly\_i; hattheta) be the conditional distribution of phi\_i for 1<=i<=N . We use the MCMC procedure used in the SAEM algorithm to estimate these conditional distributions. We empirically estimate the conditional mean E(phi\_i ly\_i; hattheta) and the conditional standard deviation sd(phi\_i ly\_i; hattheta).

# Usage

```
conddist.saemix(saemixObject,nsamp=1,max.iter=NULL,...)
```

## **Arguments**

saemixObject	an object returned by the saemix function
nsamp	Number of samples to be drawn in the conditional distribution for each subject. Defaults to 1
max.iter	Maximum number of iterations for the computation of the conditional estimates.  Defaults to twice the total number of iterations (sum(saemixObject["options"]\$nbiter.saemix)*2)
• • •	optional arguments passed to the plots. Plots will appear if the option displayProgress in the saemixObject object is TRUE

# **Details**

See PDF documentation for details of the computation. Briefly, the MCMC algorithm is used to obtain samples from the individual conditional distributions of the parameters. The algorithm is initialised for each subject to the conditional estimate of the individual parameters obtained at the end of the SAEMIX fit. A convergence criterion is used to ensure convergence of the mean and variance of the conditional distributions. When nsamp>1, several chains of the MCMC algorithm are run in parallel to obtain samples from the conditional distributions, and the convergence criterion must be achieved for all chains. When nsamp>1, the estimate of the conditional mean is obtained by averaging over the different samples.

The shrinkage for any given parameter for the conditional estimate is obtained as

Sh=1-var(eta\_i)/omega(eta)

where var(eta\_i) is the empirical variance of the estimates of the individual random effects, and omega(eta) is the estimated variance.

The function adds or modifies the following elements in the results:

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**cond.mean.phi** Conditional mean of the individual distribution of the parameters (obtained as the mean of the samples)

**cond.var.phi** Conditional variance of the individual distribution of the parameters (obtained as the mean of the estimated variance of the samples)

cond.shrinkage Estimate of the shrinkage for the conditional estimates

**cond.mean.eta** Conditional mean of the individual distribution of the parameters (obtained as the mean of the samples)

**phi.samp** An array with 3 dimensions, giving nsamp samples from the conditional distributions of the individual parameters

**phi.samp.var** The estimated individual variances for the sampled parameters phi.samp

A warning is output if the maximum number of iterations is reached without convergence (the maximum number of iterations is saemix.options\$nbiter.saemix[2]).

## Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

## References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

# See Also

SaemixData, SaemixModel, SaemixObject, saemixControl, saemix

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,
    name.group=c("Id"),name.predictors=c("Dose","Time"),
    name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
    units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")

model1cpt<-function(psi,id,xidep) {
    dose<-xidep[,1]
    tim<-xidep[,2]
    ka<-psi[id,1]
    V<-psi[id,2]
    CL<-psi[id,3]
    k<-CL/V
    ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))
    return(ypred)</pre>
```

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```
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,dimnames=list(NULL,
 c("ka", "V", "CL"))), transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),\ error.model="constant")
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# saemix.fit<-conddist.saemix(saemix.fit,nsamp=3)</pre>
# First sample from the conditional distribution
# (a N (nb of subject) by nb.etas (nb of parameters) matrix)
# saemix.fit["results"]["phi.samp"][,,1]
# Second sample
# saemix.fit["results"]["phi.samp"][,,2]
```

cow.saemix

Evolution of the weight of 560 cows, in SAEM format

## **Description**

cow. saemix contains data from winter wheat experiments.

#### Usage

cow.saemix

#### **Format**

This data frame contains the following columns:

```
cow: the id.
time: time (days).
weight: weight of the cow (kg).
birthyear: year of birth (between 1988 and 1998).
twin: existence of a twin (no=1, yes=2).
birthrank: the rank of birth (beetween 3 and 7).
```

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#### **Details**

The data used in this example is the evolution of the weight (in kg) of 560 cows. We have 9 or 10 measurements per subject.

We use an exponential growth model for this data: y\_ij = A\_i (1- B\_i exp( - K\_i t\_ij)) +epsilon\_ij

```
data(cow.saemix)
saemix.data<-saemixData(name.data=cow.saemix,header=TRUE,name.group=c("cow"),</pre>
 name.predictors=c("time"),name.response=c("weight"),
 name.covariates=c("birthyear", "twin", "birthrank"),
 units=list(x="days",y="kg",covariates=c("yr","-","-")))
growthcow<-function(psi,id,xidep) {</pre>
# input:
   psi : matrix of parameters (3 columns, a, b, k)
   id : vector of indices
   xidep: dependent variables (same nb of rows as length of id)
   a vector of predictions of length equal to length of id
 x<-xidep[,1]
 a<-psi[id,1]
 b<-psi[id,2]
 k<-psi[id,3]
 f<-a*(1-b*exp(-k*x))
 return(f)
}
saemix.model<-saemixModel(model=growthcow,</pre>
 description="Exponential growth model",
 psi0=matrix(c(700,0.9,0.02,0,0,0),ncol=3,byrow=TRUE,
 dimnames=list(NULL,c("A","B","k"))), transform.par=c(1,1,1), fixed.estim=c(1,1,1),\\
 covariate.model=matrix(c(0,0,0), ncol=3, byrow=TRUE),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),\\
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")\\
saemix.options<-list(algorithms=c(1,1,1),nb.chains=1,nbiter.saemix=c(200,100),</pre>
 seed=4526, save=FALSE, save.graphs=FALSE)
# Plotting the data
plot(saemix.data,xlab="Time (day)",ylab="Weight of the cow (kg)")
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
```

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# **Description**

These functions produce default sets of plots, corresponding to diagnostic or individual fits.

# Usage

```
default.saemix.plots(saemixObject, ...)
basic.gof(saemixObject, ...)
advanced.gof(saemixObject, ...)
covariate.fits(saemixObject, which = "parameters", ...)
individual.fits(saemixObject, ...)
```

# **Arguments**

saemixObject an object returned by the saemix function

which for covariate fits, whether they should be produced with the EBE estimates of the parameters (the default) or with random effects (which="randeff")

... optional arguments passed to the plots

#### **Details**

These functions are wrapper functions designed to produce default sets of plots to help the user assess their model fits.

## Value

by default, the following plots are produced: a plot of the data, convergence plots, plot of the likelihood by importance sampling (if computed), plots of observations versus predictions, scatterplots and distribution of residuals, boxplot of the random effects, correlations between random effects, distribution of the parameters, VPC

basic goodness-of-fit plots: convergence plots, plot of the likelihood by importance sampling (if computed), plots of observations versus predictions

advanced.gof advanced goodness-of-fit plots: scatterplots and distribution of residuals, VPC,...

covariate.fits plots of all estimated parameters versus all covariates in the dataset individual.fits

plots of individual predictions (line) overlayed on individual observations (dots) for all subjects in the dataset

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

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#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
saemix, saemix.plot.data, saemix.plot.setoptions, plot.saemix
```

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3, byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
# Reducing the number of iterations due to time constraints for CRAN
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE,nbiter.saemix=c(100,100))
saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
default.saemix.plots(saemix.fit)
# Not run (time constraints for CRAN)
# basic.gof(saemix.fit)
# Not run (time constraints for CRAN)
# advanced.gof(saemix.fit)
```

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individual.fits(saemix.fit)

fim.saemix

Computes the Fisher Information Matrix by linearisation

# **Description**

Estimate by linearisation the Fisher Information Matrix and the standard error of the estimated parameters.

## Usage

fim.saemix(saemixObject)

#### **Arguments**

saemixObject an object returned by the saemix function

#### **Details**

The inverse of the Fisher Information Matrix provides an estimate of the variance of the estimated parameters theta. This matrix cannot be computed in closed-form for nonlinear mixed-effect models; instead, an approximation is obtained as the Fisher Information Matrix of the Gaussian model deduced from the nonlinear mixed effects model after linearisation of the function f around the conditional expectation of the individual Gaussian parameters. This matrix is a block matrix (no correlations between the estimated fixed effects and the estimated variances).

# Value

The function returns an updated version of the object saemix.fit in which the following elements have been added:

**se.fixed:** standard error of fixed effects, obtained as part of the diagonal of the inverse of the Fisher Information Matrix (only when fim.saemix has been run, or when the saemix.options\$algorithms[2] is 1)

**se.omega:** standard error of the variance of random effects, obtained as part of the diagonal of the inverse of the Fisher Information Matrix (only when fim.saemix has been run, or when the saemix.options\$algorithms[2] is 1)

**se.res:** standard error of the parameters of the residual error model, obtained as part of the diagonal of the inverse of the Fisher Information Matrix (only when fim.saemix has been run, or when the saemix.options\$algorithms[2] is 1)

fim: Fisher Information Matrix

**Il.lin:** likelihood calculated by linearisation

#### Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

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#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
SaemixObject, saemix
```

```
# Running the main algorithm to estimate the population parameters
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]</pre>
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE), \ error.model="constant")
saemix.options < -list(algorithm=c(1,0,0),seed=632545,save=FALSE,save.graphs=FALSE)
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Estimating the Fisher Information Matrix using the result of saemix
# & returning the result in the same object
# fim.saemix(saemix.fit)
```

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initialize-methods

Methods for Function initialize

# **Description**

Constructor functions for Classes in the saemix package (not user-level)

#### Methods

```
signature(.Object = "SaemixData") create a SaemixData object. Please use the saemixData
    function.
signature(.Object = "SaemixModel") create a SaemixModel object Please use the saemixModel
    function.
signature(.Object = "SaemixObject") create a SaemixObject object. This object is obtained
    after a successful call to saemix
signature(.Object = "SaemixRepData") create a SaemixRepData object
signature(.Object = "SaemixRes") create a SaemixRes object
signature(.Object = "SaemixSimData") create a SaemixSimData object
```

llgq.saemix

Log-likelihood using Gaussian Quadrature

## **Description**

Estimate the log-likelihood using Gaussian Quadrature (multidimensional grid)

# Usage

```
llgq.saemix(saemixObject)
```

# **Arguments**

```
saemixObject an object returned by the saemix function
```

# **Details**

The likelihood of the observations is estimated using Gaussian Quadrature (see documentation).

# Value

the log-likelihood estimated by Gaussian Quadrature

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## Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
SaemixObject, saemix, llis. saemix
```

```
# Running the main algorithm to estimate the population parameters
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]</pre>
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]</pre>
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3,byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3, byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),\ error.model="constant")
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Estimating the likelihood by Gaussian Quadrature using the result of saemix
# & returning the result in the same object
# saemix.fit<-llgq.saemix(saemix.fit)</pre>
```

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llis.saemix

Log-likelihood using Importance Sampling

# **Description**

Estimate the log-likelihood using Importance Sampling

# Usage

```
llis.saemix(saemixObject)
```

# **Arguments**

```
saemixObject an object returned by the saemix function
```

#### **Details**

The likelihood of the observations is estimated without any approximation using a Monte-Carlo approach (see documentation).

#### Value

the log-likelihood estimated by Importance Sampling

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

# See Also

```
SaemixObject,saemix,llgq.saemix
```

```
# Running the main algorithm to estimate the population parameters
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,
    name.group=c("Id"),name.predictors=c("Dose","Time"),
    name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
    units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")</pre>
```

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```
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]</pre>
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]</pre>
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),\\
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")\\
saemix.options<-list(algorithm=c(1,0,0),seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Estimating the likelihood by importance sampling using the result of saemix
# & returning the result in the same object
# saemix.fit<-llis.saemix(saemix.fit)</pre>
```

logLik-methods

~~ Methods for Function logLik ~~

# Description

```
~~ Methods for function logLik ~~
```

## Methods

```
signature(object = "SaemixObject")
```

map.saemix

Estimates of the individual parameters (conditional mode)

## **Description**

Compute the estimates of the individual parameters PSI\_i (conditional mode - Maximum A Posteriori)

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## Usage

```
map.saemix(saemixObject)
```

#### **Arguments**

```
saemixObject an object returned by the saemix function
```

#### **Details**

The MCMC procedure is used to estimate the conditional mode (or Maximum A Posteriori) m(phi\_i lyi; hattheta) = Argmax\_phi\_i p(phi\_i lyi; hattheta)

# Value

```
saemixObject: returns the object with the estimates of the MAP parameters (see example for usage)
```

## Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
SaemixObject, saemix
```

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,
    name.group=c("Id"),name.predictors=c("Dose","Time"),
    name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
    units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")

model1cpt<-function(psi,id,xidep) {
    dose<-xidep[,1]
    tim<-xidep[,2]
    ka<-psi[id,1]
    V<-psi[id,2]
    CL<-psi[id,3]
    k<-CL/V</pre>
```

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```
ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3, byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
saemix.options<-list(algorithm=c(1,0,0), seed=632545,</pre>
 save=FALSE, save.graphs=FALSE)
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Estimating the individual parameters using the result of saemix
# & returning the result in the same object
# saemix.fit<-map.saemix(saemix.fit)</pre>
```

oxboys.saemix

Heights of Boys in Oxford

# **Description**

The oxboys. saemix data frame has 234 rows and 4 columns.

# Usage

```
oxboys.saemix
```

# **Format**

This data frame contains the following columns:

Subject: an ordered factor giving a unique identifier for each boy in the experiment

age: a numeric vector giving the standardized age (dimensionless)

**height:** a numeric vector giving the height of the boy (cm)

**Occasion:** an ordered factor - the result of converting 'age' from a continuous variable to a count so these slightly unbalanced data can be analyzed as balanced.

#### **Details**

These data are described in Goldstein (1987) as data on the height of a selection of boys from Oxford, England versus a standardized age. The dataset can be found in the package nlme.

We use an linear model for this data: y\_ij = Base\_i + slope\_i x\_ij +epsilon\_ij

PD1.saemix

## Source

Pinheiro, J. C. and Bates, D. M. (2000), \_Mixed-Effects Models in S and S-PLUS\_, Springer, New York. (Appendix A.19)

# **Examples**

```
data(oxboys.saemix)
saemix.data<-saemixData(name.data=oxboys.saemix,header=TRUE,</pre>
 name.group=c("Subject"),name.predictors=c("age"),name.response=c("height"),
 units=list(x="yr",y="cm"))
# plot the data
plot(saemix.data)
growth.linear<-function(psi,id,xidep) {</pre>
# input:
   psi : matrix of parameters (2 columns, base and slope)
   id : vector of indices
   xidep: dependent variables (same nb of rows as length of id)
   a vector of predictions of length equal to length of id
 x < -xidep[,1]
 base<-psi[id,1]
 slope<-psi[id,2]</pre>
 f<-base+slope*x
 return(f)
}
saemix.model<-saemixModel(model=growth.linear,description="Linear model",</pre>
 psi0=matrix(c(140,1),ncol=2,byrow=TRUE,dimnames=list(NULL,c("base","slope"))),
 transform.par=c(1,0), covariance.model=matrix(c(1,1,1,1), ncol=2, byrow=TRUE),
 error.model="constant")
saemix.options<-list(algorithms=c(1,1,1),nb.chains=1,seed=201004,</pre>
 save=FALSE, save.graphs=FALSE)
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
```

PD1.saemix

Data simulated according to an Emax response model, in SAEM format

# **Description**

PD1. saemix contains data from winter wheat experiments.

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## Usage

```
PD1.saemix
PD2.saemix
```

#### **Format**

This data frame contains the following columns:

```
subject: the site numberdose: simulated dose.response: simulated response.gender: gender (0 for male, 1 for female).
```

## **Details**

These examples were used by P. Girard and F. Mentre for the symposium dedicated to Comparison of Algorithms Using Simulated Data Sets and Blind Analysis, that took place in Lyon, France, September 2004.

The dataset contains 100 individuals, each receiving 3 different doses: (0, 10, 90), (5, 25, 65) or (0, 20, 30). It was assumed that doses were given in a cross-over study with sufficient wash out period to avoid carry over. Responses (y\_ij) were simulated with the following pharmacodynamic model:

```
y_{ij} = E0_i + D_{ij} Emax_i/(D_{ij} + ED50_i) + epsilon_{ij}
```

The individual parameters were simulated according to

```
log (E0_i) = log (E0) + eta_i 1 log (Emax_i) = log (Emax) + eta_i 2 log (E50_i) = log (E50) + beta w_i + eta_i 3
```

PD1.saemix contains the data simulated with a gender effect, beta=0.3.

PD2.saemix contains the data simulated without a gender effect, beta=0.

# Source

Girard P., Mentre F. Comparison of Algorithms Using Simulated Data Sets and Blind Analysis workshop, Lyon, France, September 2004.

```
data(PD1.saemix)
saemix.data<-saemixData(name.data=PD1.saemix,header=TRUE,name.group=c("subject"),
    name.predictors=c("dose"),name.response=c("response"),
    name.covariates=c("gender"), units=list(x="mg",y="-",covariates=c("-")))

modelemax<-function(psi,id,xidep) {
    # input:
    # psi : matrix of parameters (3 columns, E0, Emax, EC50)
# id : vector of indices</pre>
```

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```
xidep : dependent variables (same nb of rows as length of id)
   a vector of predictions of length equal to length of id
 dose<-xidep[,1]</pre>
 e0<-psi[id,1]
 emax<-psi[id,2]
 e50<-psi[id,3]
 f<-e0+emax*dose/(e50+dose)
 return(f)
}
# Plotting the data
plot(saemix.data,main="Simulated data PD1")
# Not run (strict time constraints for CRAN)
# Compare models with and without covariates with LL by Importance Sampling
# SE not computed
model1<-saemixModel(model=modelemax,description="Emax growth model",</pre>
 psi0=matrix(c(20,300,20,0,0,0),ncol=3,byrow=TRUE,dimnames=list(NULL,
 c("E0", "Emax", "EC50"))), transform.par=c(1,1,1),
 covariate.model=matrix(c(0,0,0), ncol=3,byrow=TRUE),fixed.estim=c(1,1,1))
model2<-saemixModel(model=modelemax,description="Emax growth model",</pre>
 psi0=matrix(c(20,300,20,0,0,0),ncol=3,byrow=TRUE,dimnames=list(NULL,
 c("E0","Emax","EC50"))), transform.par=c(1,1,1),
 covariate.model=matrix(c(0,0,1), ncol=3,byrow=TRUE),fixed.estim=c(1,1,1))
saemix.options<-list(algorithms=c(0,1,1),nb.chains=3,seed=765754,</pre>
 nbiter.saemix=c(500,300),save=FALSE,save.graphs=FALSE)
# Not run (strict time constraints for CRAN)
# fit1<-saemix(model1, saemix.data, saemix.options)</pre>
# fit2<-saemix(model2, saemix.data, saemix.options)</pre>
# wstat<-(-2)*(fit1["results"]["ll.is"]-fit2["results"]["ll.is"])</pre>
# cat("LRT test for covariate effect on EC50: p-value=",1-pchisq(wstat,1),"\n")
```

plot-methods

Methods for Function plot

#### Description

Methods for function plot

# Methods

```
signature(x = "ANY") default plot function ?
signature(x = "SaemixData") Plots the data. Defaults to a spaghetti plot of response versus
predictor, with lines joining the data for one individual.
```

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```
signature(x = "SaemixModel") Plots prediction of the model
signature(x = "SaemixObject") This method gives access to a number of plots that can be
performed on a SaemixObject
signature(x = "SaemixSimData") Plots simulated datasets
```

predict-methods

Methods for Function predict

# **Description**

Methods for function predict

## Methods

```
signature(object = "ANY") Default predict functions
signature(object = "SaemixObject") Computes predictions using the results of an SAEM fit
```

print-methods

Methods for Function print

# Description

Prints a summary of an object

#### Methods

```
signature(x = "ANY") Default print function
signature(x = "SaemixData") Prints a summary of a SaemixData object
signature(x = "SaemixModel") Prints a summary of a SaemixModel object
signature(x = "SaemixObject") Prints a summary of the results from a SAEMIX fit
signature(x = "SaemixRes") Not user-level
```

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psi	Functions to extract the individual estimates of the parameters and
	random effects

# **Description**

These three functions are used to access the estimates of individual parameters and random effects.

## Usage

```
psi(object, indiv.par)
phi(object, indiv.par)
eta(object, indiv.par)
```

# Arguments

object an object returned by the saemix function

indiv.par a string giving the type of estimate to be used (one of "map", for the Maximum

A Posteriori estimate, or "eap", for conditional estimate). Defaults to "map"

## **Details**

The psi\_i represent the individual parameter estimates. In the SAEM algorithm, these parameters are assumed to be a transformation of a Gaussian random vector phi\_i, where the phi\_i can be written as a function of the individual random effects (eta\_i), the covariate matrix (C\_i) and the vector of fixed effects (mu):

```
phi_i = C_i mu + eta_i
```

More details can be found in the PDF documentation.

# Value

These functions are used to access and output the estimates of parameters and random effects. When the object passed to the function does not contain these estimates, they are automatically computed. The object is then returned (invisibly) with these estimates added to the results.

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

# References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

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## See Also

SaemixData, SaemixModel, SaemixObject, saemixControl, plot. saemix

# **Examples**

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0), ncol=3, byrow=TRUE), fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3, byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
saemix.options<-list(algorithm=c(1,0,0),seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# psi(saemix.fit)
# phi(saemix.fit)
# eta(saemix.fit,indiv.par="eap")
```

psi-methods

Methods for Functions psi, phi and eta

# **Description**

These methods are used to access estimates of individual parameters and random effects

## Methods

signature(object = "SaemixObject") please refer to the PDF documentation for the models

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saemix	Stochastic Approximation Expectation Maximization (SAEM) algorithm

# **Description**

SAEM algorithm perform parameter estimation for nonlinear mixed effects models without any approximation of the model (linearization, quadrature approximation, . . . )

# Usage

```
saemix(model, data, control = list())
```

# **Arguments**

model	an object of class SaemixModel, created by a call to the function saemixModel
data	an object of class SaemixData, created by a call to the function saemixData
control	a list of options, see saemixControl

#### **Details**

The SAEM algorithm is a stochastic approximation version of the standard EM algorithm proposed by Kuhn and Lavielle (see reference). Details of the algorithm can be found in the pdf file accompanying the package.

## Value

An object of class SaemixObject containing the results of the fit of the data by the non-linear mixed effect model. A summary of the results is printed out to the terminal, and, provided the appropriate options have not been changed, numerical and graphical outputs are saved in a directory.

## Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

## References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

# See Also

SaemixData,SaemixModel, SaemixObject, saemixControl, plot.saemix

## **Examples**

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
  name.group=c("Id"),name.predictors=c("Dose","Time"),
  name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
  units=list(x="hr",y="mg/L", covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]</pre>
 V<-psi[id,2]</pre>
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")\\
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,list(seed=632545,directory="newtheo",</pre>
# save=FALSE, save.graphs=FALSE))
# Prints a summary of the results
# print(saemix.fit)
# Outputs the estimates of individual parameters
# psi(saemix.fit)
# Shows some diagnostic plots to evaluate the fit
# plot(saemix.fit)
```

saemix.plot.data

Functions implementing each type of plot in SAEM

## **Description**

Several plots (selectable by the type argument) are currently available: convergence plot, individual plots, predictions versus observations, distribution plots, VPC, residual plots.

#### **Usage**

```
saemix.plot.data(saemixObject, ...)
saemix.plot.convergence(saemixObject,niter=0, ...)
saemix.plot.llis(saemixObject, ...)
saemix.plot.obsvspred(saemixObject, ...)
saemix.plot.distribresiduals(saemixObject, ...)
saemix.plot.scatterresiduals(saemixObject, ...)
saemix.plot.fits(saemixObject, ...)
saemix.plot.distpsi(saemixObject, ...)
saemix.plot.randeff(saemixObject, ...)
saemix.plot.correlations(saemixObject, ...)
saemix.plot.parcov(saemixObject, ...)
saemix.plot.randeffcov(saemixObject, ...)
saemix.plot.npde(saemixObject, ...)
saemix.plot.vpc(saemixObject,npc = FALSE, ...)
saemix.plot.parcov.aux(saemixObject, partype = "p", ...)
compute.sres(saemixObject)
compute.eta.map(saemixObject)
```

#### **Arguments**

an object returned by the saemix function

optional arguments passed to the plots

npc for VPC, computes Numerical Predictive Checks (currently not implemented)

niter the convergence plots are shown up to iteration "niter". Defaults to 0, which indicates the convergence plots should be plotted up to the maximal number of iterations set for the algorithm

partype (this function is not user-level)

#### **Details**

These functions implement plots different graphs related to the algorithm (convergence plots, likelihood estimation) as well as diagnostic graphs. A description is provided in the PDF documentation. saemix.plot.parcov.aux, compute.sres and compute.eta.map are helper functions, not intended to be called by the user directly.

By default, the following plots are produced:

**saemix.plot.data:** A spaghetti plot of the data, displaying the observed data y as a function of the regression variable (time for a PK application)

**saemix.plot.convergence:** For each parameter in the model, this plot shows the evolution of the parameter estimate versus the iteration number

**saemix.plot.llis:** Graph showing the evolution of the log-likelihood during the estimation by importance sampling

**saemix.plot.obsvspred:** Plot of the predictions computed with the population parameters versus the observations (left), and plot of the predictions computed with the individual parameters versus the observations (right)

**saemix.plot.scatterresiduals:** Scatterplot of the residuals versus the predictor (top) and versus predictions (bottom), for weighted residuals (population residuals, left), individual weighted residuals (middle) and npde (right).

- **saemix.plot.distribresiduals:** Distribution of the residuals, plotted as histogram (top) and as a QQ-plot (bottom), for weighted residuals (population residuals, left), individual weighted residuals (middle) and npde (right).
- saemix.plot.fits: Model fits. Individual fits are obtained using the individual parameters with the individual covariates. Population fits are obtained using the population parameters with the individual covariates (red) and the individual parameters with the individual covariates (green). By default the individual plots are displayed.
- saemix.plot.distpsi: Distribution of the parameters (conditional on covariates when some are included in the model). A histogram of individual parameter estimates can be overlayed on the plot, but it should be noted that the histogram does not make sense when there are covariates influencing the parameters and a warning will be displayed

saemix.plot.randeff: Boxplot of the random effects

saemix.plot.correlations: Correlation between the random effects

- **saemix.plot.parcov:** Plots of the estimates of the individual parameters versus the covariates, using scatterplot for continuous covariates, boxplot for categorical covariates
- **saemix.plot.randeffcov:** Plots of the estimates of the random effects versus the covariates, using scatterplot for continuous covariates, boxplot for categorical covariates
- **saemix.plot.npde:** Plots 4 graphs to evaluate the shape of the distribution of the normalised prediction distribution errors (npde)
- **saemix.plot.vpc:** Visual Predictive Check, with options to include the prediction intervals around the boundaries of the selected interval as well as around the median (50th percentile of the simulated data). Several methods are available to define binning on the X-axis (see methods in the PDF guide).

Each plot can be customised by modifying options, either through a list of options set by the saemix.plot.setoptions function, or on the fly by passing an option in the call to the plot (see examples).

# Value

None

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

# References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

## See Also

SaemixObject,saemix.plot.setoptions, saemix.plot.select, plot.saemix

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
  return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0), ncol=3, byrow=TRUE), fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
saemix.options<-list(seed=632545, save=FALSE, save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Simulate data and compute weighted residuals and npde
# saemix.fit<-compute.sres(saemix.fit)</pre>
# saemix.plot.data(saemix.fit)
# Convergence
# saemix.plot.convergence(saemix.fit)
# Individual plot for subject 1, smoothed
# saemix.plot.fits(saemix.fit,ilist=1,smooth=TRUE)
# Individual plot for subject 1 to 12, with ask set to TRUE
# (the system will pause before a new graph is produced)
# saemix.plot.fits(saemix.fit,ilist=c(1:12),ask=TRUE)
```

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```
# Diagnostic plot: observations versus population predictions
# par(mfrow=c(1,1))
# saemix.plot.obsvspred(saemix.fit,level=0,new=FALSE)
# LL by Importance Sampling
# saemix.plot.llis(saemix.fit)
# Scatter plot of residuals
# saemix.plot.scatterresiduals(saemix.fit)
# Boxplot of random effects
# saemix.plot.randeff(saemix.fit)
# Relationships between parameters and covariates
# saemix.plot.parcov(saemix.fit)
# Relationships between parameters and covariates, on the same page
# par(mfrow=c(3,2))
# saemix.plot.parcov(saemix.fit,new=FALSE)
# VPC, default options (10 bins, equal number of observations in each bin)
# Not run (time constraints for CRAN)
# saemix.plot.vpc(saemix.fit)
# VPC, user-defined breaks for binning
# Not run (time constraints for CRAN)
# saemix.plot.vpc(saemix.fit,vpc.method="user", vpc.breaks=c(0.4,0.8,1.5,2.5,4,5.5,8,10,13))
```

saemix.plot.select

Plots of the results obtained by SAEM

# **Description**

Several plots (selectable by the type argument) are currently available: convergence plot, individual plots, predictions versus observations, distribution plots, residual plots, VPC.

# Usage

```
saemix.plot.select(saemixObject, data = FALSE, convergence = FALSE,
  likelihood = FALSE, individual.fit = FALSE, population.fit = FALSE,
  both.fit = FALSE, observations.vs.predictions = FALSE,
  residuals.scatter = FALSE, residuals.distribution = FALSE,
  random.effects = FALSE, correlations = FALSE,
  parameters.vs.covariates = FALSE, randeff.vs.covariates = FALSE,
  marginal.distribution = FALSE, vpc = FALSE, npde = FALSE, ...)
```

# **Arguments**

saemixObject an object returned by the saemix function

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data if TRUE, produce a plot of the data. Defaults to FALSE convergence if TRUE, produce a convergence plot. Defaults to FALSE

likelihood if TRUE, produce a plot of the estimation of the LL by importance sampling.

Defaults to FALSE

individual.fit if TRUE, produce individual fits with individual estimates. Defaults to FALSE population.fit if TRUE, produce individual fits with population estimates. Defaults to FALSE both.fit if TRUE, produce individual fits with both individual and population estimates.

Defaults to FALSE

observations.vs.predictions

if TRUE, produce a plot of observations versus predictions. Defaults to FALSE

residuals.scatter

if TRUE, produce scatterplots of residuals versus predictor and predictions. Defaults to FALSE

residuals.distribution

if TRUE, produce plots of the distribution of residuals. Defaults to FALSE

random.effects if TRUE, produce boxplots of the random effects. Defaults to FALSE

correlations if TRUE, produce a matrix plot showing the correlation between random effects.

Defaults to FALSE

parameters.vs.covariates

if TRUE, produce plots of the relationships between parameters and covariates, using the Empirical Bayes Estimates of individual parameters. Defaults to FALSE

randeff.vs.covariates

if TRUE, produce plots of the relationships between random effects and covariates, using the Empirical Bayes Estimates of individual random effects. Defaults to FALSE

marginal.distribution

if TRUE, produce plots of the marginal distribution of the random effects. Defaults to FALSE

vpc if TRUE, produce Visual Predictive Check plots. Defaults to FALSE

npde if TRUE, produce plots of the npde. Defaults to FALSE

... optional arguments passed to the plots

# Details

This function plots different graphs related to the algorithm (convergence plots, likelihood estimation) as well as diagnostic graphs. A description is provided in the PDF documentation.

**data** A spaghetti plot of the data, displaying the observed data y as a function of the regression variable (eg time for a PK application)

**convergence** For each parameter in the model, this plot shows the evolution of the parameter estimate versus the iteration number

**likelihood** Estimation of the likelihood estimated by importance sampling, as a function of the number of MCMC samples

individual.fit Individual fits, using the individual parameters with the individual covariates

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population.fit Individual fits, using the population parameters with the individual covariates

**both.fit** Individual fits, using the population parameters with the individual covariates and the individual parameters with the individual covariates

**observations.vs.predictions** Plot of the predictions computed with the population parameters versus the observations (left), and plot of the predictions computed with the individual parameters versus the observations (right)

**residuals.scatter** Scatterplot of standardised residuals versus the X predictor and versus predictions. These plots are shown for individual and population residuals, as well as for npde when they are available

**residuals.distribution** Distribution of standardised residuals, using histograms and QQ-plot. These plots are shown for individual and population residuals, as well as for npde when they are available

random.effects Boxplot of the random effects

correlations Correlation between the random effects, with a smoothing spline

**parameters.versus.covariates** Plots of the estimate of the individual parameters versus the covariates, using scatterplot for continuous covariates, boxplot for categorical covariates

**randeff.versus.covariates** Plots of the estimate of the individual random effects versus the covariates, using scatterplot for continuous covariates, boxplot for categorical covariates

**marginal.distribution** Distribution of each parameter in the model (conditional on covariates when some are included in the model)

npde Plot of npde as in package npde

vpc Visual Predictive Check

#### Value

None

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

# References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
SaemixObject,saemix, default.saemix.plots, saemix.plot.setoptions, saemix.plot.data, saemix.plot.convergence, saemix.plot.llis, saemix.plot.randeff, saemix.plot.obsvspred, saemix.plot.fits, saemix.plot.parcov, saemix.plot.randeffcov, saemix.plot.distpsi, saemix.plot.scatterresiduals, saemix.plot.distribresiduals, saemix.plot.vpc
```

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# **Examples**

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 \label{list} units=list(x="hr",y="mg/L",covariates=c("kg","-")), \ name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]</pre>
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]</pre>
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))), transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0), ncol=3, byrow=TRUE), fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3, byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")\\
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# saemix.plot.select(saemix.fit,data=TRUE,main="Spaghetti plot of data")
# Putting several graphs on the same plot
# par(mfrow=c(2,2))
# saemix.plot.select(saemix.fit,data=TRUE,vpc=TRUE,observations.vs.predictions=TRUE, new=FALSE)
```

saemix.plot.setoptions

Function setting the default options for the plots in SAEM

# **Description**

This function can be used to create a list containing the default options and arguments used by the plot functions.

#### Usage

```
saemix.plot.setoptions(saemixObject)
saemix.data.setoptions(saemix.data)
replace.plot.options(plot.opt, ...)
replace.data.options(plot.opt, ...)
```

# **Arguments**

saemixObject an object returned by the saemix function
saemix.data an SaemixData object returned by the read.saemixData function
plot.opt current graphic options
optional arguments passed to the plots, which will be used to override the current

options defined in plot.opt

#### **Details**

A more detailed description of the options set via these lists is provided in the PDF documentation. The "replace" functions are helper functions used within the plot functions. saemix.plot.setoptions has more available options than saemix.data.setoptions since it applies to all possible plots while the latter only applies to data.

ablinecol Color of the lines added to the plots (default: "DarkRed")

**ablinelty** Type of the lines added to the plots. Defaults to 2 (dashed line)

**ablinelwd** Width of the lines added to the plots (default: 2)

ask A logical value. If TRUE, users will be prompted before each new plot. Defaults to FALSE

**cex** A numerical value giving the amount by which plotting text and symbols should be magnified relative to the default. Defaults to 1 (no magnification)

**cex.axis** Magnification to be used for axis annotation relative to the current setting of 'cex'. Defaults to 1 (no magnification)

**cex.main** Magnification to be used for main titles relative to the current setting of 'cex'. Defaults to 1 (no magnification)

**cex.lab** Magnification to be used for x and y labels relative to the current setting of 'cex'. Defaults to 1 (no magnification)

**col.fillmed** For the VPC plots: color filling the prediction interval for the median. Defaults to "pink"

**col.fillpi** For the VPC plots: color filling the prediction interval for the limits of the prediction interval. Defaults to "slategray1"

**col.lmed** For the VPC plots: color of the line showing the median of the simulated data. Defaults to "indianred4"

**col.lobs** For the VPC plots: color of the lines showing the median, 2.5 and 97.5th percentiles (for a 95

**col.lpi** For the VPC plots: color of the line showing the boundaries of the prediction intervals. Defaults to "slategray4"

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col.obs For the VPC plots: color used to plot the observations. Defaults to "steelblue4"

**cov.name** Name of the covariate to be used in the plots. Defaults to the first covariate in the model

**cov.value** Value of the covariate to be used in the plots. Defaults to NA, indicating that the median value of the covariate (for continuous covariates) or the reference category (for categorical covariates) will be used

ilist List of indices of subjects to be included in the individual plots (defaults to all subjects)

**indiv.par** a string, giving the type of the individual estimates ("map"= conditional mode, "eap"=conditional mean). Defaults to conditional mode

**lcol** Main line color (default: black)

line.smooth Type of smoothing when a smoothed line is used in the plot ("m": mean value, "l": linear regression; "s": natural splines). Several options may be combined, for instance "ls" will add both a linear regression line and a line representing the fit of a natural spline. Defaults to "s"

lty Line type. Defaults to 1, corresponding to a straight line

**Ity.lmed** For the VPC plots: type of the line showing the median of the simulated data. Defaults to 2 (dashed)

lty.obs For the VPC plots: type of the line showing the observed data. Defaults to 1

**Ity.lpi** For the VPC plots: type of the line showing the boundaries of the simulated data. Defaults to 2 (dashed)

**lwd** Line width (default: 1)

**lwd.lmed** For the VPC plots: thickness of the line showing the median of the simulated data. Defaults to 2

**lwd.obs** For the VPC plots: thickness of the line showing the median and boundaries of the observed data. Defaults to 2

**lwd.lpi** For the VPC plots: thickness of the line showing the boundaries of the simulated data. Defaults to 1

**par.name** Name of the parameter to be used in the plots. Defaults to the first parameter in the model

**pch** Symbol type. Defaults to 20, corresponding to small dots

pcol Main symbol color (default: black)

**range** Range (expressed in number of SD) over which to plot the marginal distribution. Defaults to 4, so that the random effects for the marginal distribution is taken over the range [-4 SD; 4 SD]

**res.plot** Type of residual plot ("res.vs.x": scatterplot versus X, "res.vs.pred": scatterplot versus predictions, "hist": histogram, "qqplot": QQ-plot) (default: "res.vs.x")

**smooth** When TRUE, smoothed lines are added in the plots of predictions versus observations (default: FALSE)

tit Title of the graph (default: none)

**type** Type of the plot (as in the *R* plot function. Defaults to "b", so that both lines and symbols are shown

**units** Name of the predictor used in the plots (X). Defaults to the name of the first predictor in the model (saemix.data\$names\$predictors[1])

**vpc.bin** Number of binning intervals when plotting the VPC (the (vpc.bin-1) breakpoints are taken as the empirical quantiles of the X data). Defaults to 10

**vpc.interval** Size of the prediction intervals.Defaults to 0.95 for the 95% prediction interval **vpc.obs** Should the observations be overlayed on the VPC plot. Defaults to TRUE

vpc.pi Should prediction bands be computed around the median and the bounds of the prediction intervals for the VPC. Defaults to TRUE

**xlab** Label for the X-axis. Defaults to the name of the X predictor followed by the unit in bracket (eg "Time (hr)")

**xlim** Range for the X-axis. Defaults to NA, indicating that the range is to be set by the plot function

xlog A logical value. If TRUE, a logarithmic scale is in use. Defaults to FALSE

**xname** Name of the predictor used in the plots (X)

**ylab** Label for the Y-axis. Defaults to the name of the response followed by the unit in bracket (eg "Concentration (mg/L)" (Default: none)

ylim Range for the Y-axis. Defaults to NA, indicating that the range is to be set by the plot function

ylog A logical value. If TRUE, a logarithmic scale is in use. Defaults to FALSE

Plotting a SaemixData object also allows the following options:

**individual** if TRUE, plots separate plots for each individual, otherwise plots a spaghetti plot of all the data. Defaults to FALSE

**limit** for individual plots, plots only a limited number of subjets (nmax). Defaults to TRUE **nmax** for individual plots, when limit is TRUE, the maximum number of plots to produce. Defaults to 12

**sample** for individual plots, if TRUE, randomly samples nmax different subjects to plot. Defaults to FALSE (the first nmax subjects are used in the plots)

# Value

A list containing the options set at their default value. This list can be stored in an object and its elements modified to provide suitable graphs.

## Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

## References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

SaemixObject,saemix.plot.data,saemix.plot.convergence,saemix.plot.llis,saemix.plot.randeff,saemix.plot.obsvspred,saemix.plot.fits,saemix.plot.parcov,saemix.plot.distpsi,saemix.plot.scatterressaemix.plot.vpc

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## **Examples**

```
# Theophylline example, after a call to fit.saemix (see examples)
# Not run
# sopt<-saemix.plot.setoptions(saemix.fit)
# sopt$ask<-TRUE</pre>
```

saemix.plots

General plot function from SAEM

# **Description**

Several plots (selectable by the type argument) are currently available: convergence plot, individual plots, predictions versus observations, distribution plots, VPC, residual plots.

# Usage

```
plot(x,y, ...)
```

# **Arguments**

x an object returned by the saemix function

y empty

... optional arguments passed to the plots

# Details

This is the generic plot function for an SaemixObject object, which implements different graphs related to the algorithm (convergence plots, likelihood estimation) as well as diagnostic graphs. A description is provided in the PDF documentation. Arguments such as main, xlab, etc... that can be given to the generic plot function may be used, and will be interpreted according to the type of plot that is to be drawn.

A special argument plot.type can be set to determine the type of plot; it can be one of:

**data:** A spaghetti plot of the data, displaying the observed data y as a function of the regression variable (time for a PK application)

**convergence:** For each parameter in the model, this plot shows the evolution of the parameter estimate versus the iteration number

**likelihood:** Graph showing the evolution of the log-likelihood during the estimation by importance sampling

**observations.vs.predictions:** Plot of the predictions computed with the population parameters versus the observations (left), and plot of the predictions computed with the individual parameters versus the observations (right)

**residuals.scatter:** Scatterplot of the residuals versus the predictor (top) and versus predictions (bottom), for weighted residuals (population residuals, left), individual weighted residuals (middle) and npde (right).

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**residuals.distribution:** Distribution of the residuals, plotted as histogram (top) and as a QQ-plot (bottom), for weighted residuals (population residuals, left), individual weighted residuals (middle) and npde (right).

**individual.fit:** Individual fits are obtained using the individual parameters with the individual covariates

**population.fit:** Population fits are obtained using the population parameters with the individual covariates

**both.fit:** Individual fits, superposing fits obtained using the population parameters with the individual covariates (red) and using the individual parameters with the individual covariates (green)

marginal.distribution: Distribution of the parameters (conditional on covariates when some are included in the model). A histogram of individual parameter estimates can be overlayed on the plot, but it should be noted that the histogram does not make sense when there are covariates influencing the parameters and a warning will be displayed

random.effects: Boxplot of the random effects

correlations: Correlation between the random effects

**parameters.vs.covariates:** Plots of the estimates of the individual parameters versus the covariates, using scatterplot for continuous covariates, boxplot for categorical covariates

**randeff.vs.covariates:** Plots of the estimates of the random effects versus the covariates, using scatterplot for continuous covariates, boxplot for categorical covariates

**npde:** Plots 4 graphs to evaluate the shape of the distribution of the normalised prediction distribution errors (npde)

**vpc:** Visual Predictive Check, with options to include the prediction intervals around the boundaries of the selected interval as well as around the median (50th percentile of the simulated data).

In addition, the following values for plot.type produce a series of plots:

**reduced:** produces the following plots: plot of the data, convergence plots, plot of the likelihood by importance sampling (if computed), plots of observations versus predictions. This is the default behaviour of the plot function applied to an SaemixObject object

**full:** produces the following plots: plot of the data, convergence plots, plot of the likelihood by importance sampling (if computed), plots of observations versus predictions, scatterplots and distribution of residuals, VPC, npde, boxplot of the random effects, distribution of the parameters, correlations between random effects, plots of the relationships between individually estimated parameters and covariates, plots of the relationships between individually estimated random effects and covariates

Each plot can be customised by modifying options, either through a list of options set by the saemix.plot.setoptions function, or on the fly by passing an option in the call to the plot (see examples).

# Value

None

#### Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

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#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
SaemixObject, saemix, saemix.plot.setoptions, saemix.plot.select, saemix.plot.data
```

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Set of default plots
# plot(saemix.fit)
# plot(saemix.fit,plot.type="data")
# Convergence
# plot(saemix.fit,plot.type="convergence")
```

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```
# Individual plot for subject 1, smoothed
# plot(saemix.fit,plot.type="individual.fit",ilist=1,smooth=TRUE)
# Individual plot for subject 1 to 12, with ask set to TRUE
# (the system will pause before a new graph is produced)
# plot(saemix.fit,plot.type="individual.fit",ilist=c(1:12),ask=TRUE)
# Diagnostic plot: observations versus population predictions
# par(mfrow=c(1,1))
# plot(saemix.fit,plot.type="observations.vs.predictions",level=0,new=FALSE)
# LL by Importance Sampling
# plot(saemix.fit,plot.type="likelihood")
# Scatter plot of residuals
# Data will be simulated to compute weighted residuals and npde
# the results shall be silently added to the object saemix.fit
# plot(saemix.fit,plot.type="residuals.scatter")
# Boxplot of random effects
# plot(saemix.fit,plot.type="random.effects")
# Relationships between parameters and covariates
# plot(saemix.fit,plot.type="parameters.vs.covariates")
# Relationships between parameters and covariates, on the same page
# par(mfrow=c(3,2))
# plot(saemix.fit,plot.type="parameters.vs.covariates",new=FALSE)
# VPC
# Not run (time constraints for CRAN)
# plot(saemix.fit,plot.type="vpc")
```

saemixControl

List of options for running the algorithm SAEM

## **Description**

List containing the variables relative to the optimisation algorithm. All these elements are optional and will be set to default values when running the algorithm if they are not specified by the user.

# Usage

```
saemixControl(algorithms = c(1, 1, 1), nbiter.saemix = c(300, 100),
nb.chains = 1, fix.seed = TRUE, seed = 23456, nmc.is = 5000, nu.is = 4,
print.is = FALSE, nbdisplay = 100, displayProgress = TRUE, nbiter.burn = 5,
nbiter.mcmc = c(2, 2, 2), proba.mcmc = 0.4, stepsize.rw = 0.4, rw.init = 0.5,
alpha.sa = 0.97, nnodes.gq = 12, nsd.gq = 4, maxim.maxiter = 100,
nb.sim = 1000, nb.simpred = 100, ipar.lmcmc = 50, ipar.rmcmc = 0.05,
```

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```
print = TRUE, save = TRUE, save.graphs = TRUE, directory = "newdir",
warnings = FALSE)
```

## **Arguments**

A number of variables relative to the optimisation algorithm can be set before running the SAEM algorithm; all options not set by the user are given default values as indicated below:

a vector of 1s specifying which algorithms are to be run. Defaults to c(1,1,1) (respectively estimation of the individual parameters (MAP estimates), estimation of the Fisher Information Matrix and log-likelihood by linearisation, and estimation of the log-likelihood by importance sampling)

abgoeitbasmix nb of iterations in each step (a vector containing 2 elements)

nb. chains nb of chains to be run in parallel in the MCMC algorithm. Defaults to 1.

nbiter.burn nb of iterations for burning

nbiter.mcmc nb of iterations in each kernel during the MCMC step

proba.mcmc probability of acceptance

stepsize.rw stepsize for kernels q2 and q3. Defaults to 0.4

rw.init initial variance parameters for kernels. Defaults to 0.5

alpha.sa parameter controlling cooling in the Simulated Annealing algorithm. Defaults

to 0.97

fix.seed TRUE (default) to use a fixed seed for the random number generator. When

FALSE, the random number generator is initialised using a new seed, created from the current time. Hence, different sessions started at (sufficiently) different times will give different simulation results. The seed is stored in the element

seed of the options list.

seed for the random number generator. Defaults to 123456

nmc.is nb of samples used when computing the likelihood through importance sampling

nu.is number of degrees of freedom of the Student distribution used for the estimation

of the log-likelihood by Importance Sampling. Defaults to 4

print.is when TRUE, a plot of the likelihood as a function of the number of MCMC sam-

ples when computing the likelihood through importance sampling is produced

and updated every 500 samples. Defaults to FALSE

nbdisplay nb of iterations after which to display progress

displayProgress

when TRUE, the convergence plots are plotted after every nbdisplay iteration, and a dot is written in the terminal window to indicate progress. When FALSE,

plots are not shown and the algorithm runs silently. Defaults to TRUE

nnodes.gq number of nodes to use for the Gaussian quadrature when computing the likeli-

hood with this method (defaults to 12)

nsd.gq span (in SD) over which to integrate when computing the likelihood by Gaussian

quadrature. Defaults to 4 (eg 4 times the SD)

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maxim.maxiter	Maximum number of iterations to use when maximising the fixed effects in the algorithm. Defaults to $100$
nb.sim	number of simulations to perform to produce the VPC plots or compute npde. Defaults to 1000
nb.simpred	number of simulations used to compute mean predictions (ypred element), taken as a random sample within the nb.sim simulations used for npde
ipar.lmcmc	number of iterations required to assume convergence for the conditional estimates. Defaults to $50$
ipar.rmcmc	confidence interval for the conditional mean and variance. Defaults to 0.95
print	whether the results of the fit should be printed out. Defaults to TRUE
save	whether the results of the fit should be saved to a file. Defaults to TRUE
save.graphs	whether diagnostic graphs and individual graphs should be saved to files. Defaults to TRUE
directory	the directory in which to save the results. Defaults to "newdir" in the current directory
warnings	whether warnings should be output during the fit. Defaults to FALSE

## **Details**

All the variables are optional and will be set to their default value when running saemix.

The function saemix returns an object with an element options containing the options used for the algorithm, with defaults set for elements which have not been specified by the user.

These elements are used in subsequent functions and are not meant to be used directly.

## Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

# References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

# See Also

SaemixData,SaemixModel, SaemixObject, saemix

```
# All default options
saemix.options<-saemixControl()
# All default options, changing seed
saemix.options<-saemixControl(seed=632545)</pre>
```

saemixData 43

|--|

# Description

This function creates a SaemixData object. The only mandatory argument is the name of the dataset. If the dataset has a header (or named columns), the program will attempt to detect which column correspond to ID, predictor(s) and response. Warning messages will be printed during the object creation and should be read for details.

# Usage

```
saemixData(name.data, header, sep, na, name.group, name.predictors,
  name.response, name.X, name.covariates = c(),
  units = list(x = "", y = "", covariates = c()))
```

# Arguments

name.data	name of the dataset (can be a character string giving the name of a file on disk or of a dataset in the R session, or the name of a dataset
header	whether the dataset/file contains a header. Defaults to TRUE
sep	the field separator character. Defaults to any number of blank spaces ("")
na	a character vector of the strings which are to be interpreted as NA values. Defaults to $c(NA)$
name.group	name (or number) of the column containing the subject id
name.predictors	
	name (or number) of the column(s) containing the predictors (the algorithm requires at least one predictor $\boldsymbol{x})$
name.response	name (or number) of the column containing the response variable y modelled by $predictor(s) \ x$
name.covariates	
	name (or number) of the $column(s)$ containing the covariates, if present (otherwise missing)
name.X	name of the column containing the regression variable to be used on the $\boldsymbol{X}$ axis in the plots (defaults to the first predictor)
units	list with up to three elements, $x$ , $y$ and optionally covariates, containing the units for the $X$ and $Y$ variables respectively, as well as the units for the different covariates (defaults to empty)

# **Details**

This function is the user-friendly constructor for the SaemixData object class. The read.saemixData is a helper function, used to read the dataset, and is not intended to be called directly.

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# Value

A SaemixData object (see saemixData).

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

# References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

SaemixData, SaemixModel, saemixControl, saemix

#### **Examples**

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,
    name.group=c("Id"),name.predictors=c("Dose","Time"),
    name.response=c("Concentration"),name.covariates=c("Weight","Sex"),
    units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
print(saemix.data)
plot(saemix.data)</pre>
```

SaemixData-class

Class "SaemixData"

# **Description**

An object of the SaemixData class, representing a longitudinal data structure, used by the SAEM algorithm.

# **Objects from the Class**

An object of the SaemixData class can be created by using the function saemixData

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#### Slots

name.data: Object of class "character": name of the dataset
header: Object of class "logical": whether the dataset/file contains a header. Defaults to TRUE
sep: Object of class "character": the field separator character

na: Object of class "character": a character vector of the strings which are to be interpreted as NA values

name.group: Object of class "character": name of the column containing the subject id

name.predictors: Object of class "character": name of the column(s) containing the predictors

name.response: Object of class "character": name of the column containing the response variable y modelled by predictor(s) x

name.covariates: Object of class "character": name of the column(s) containing the covariates, if present (otherwise empty)

name.X: Object of class "character": name of the column containing the regression variable to be used on the X axis in the plots

units: Object of class "list": list with up to three elements, x, y and optionally covariates, containing the units for the X and Y variables respectively, as well as the units for the different covariates

data: Object of class "data.frame": dataframe containing the data, with columns for id (name.group), predictors (name.predictors), response (name.response), and covariates if present in the dataset (name.covariates). A column "index" contains the subject index (used to map the subject id). The column names, except for the additional column index, correspond to the names in the original dataset.

N: Object of class "numeric": number of subjects

yorig: Object of class "numeric": response data, on the original scale. Used when the error model is exponential

ntot.obs: Object of class "numeric": total number of observations

nind.obs: Object of class "numeric": vector containing the number of observations for each subject

```
[<- signature(x = "SaemixData"): replace elements of object
[ signature(x = "SaemixData"): access elements of object
initialize signature(.Object = "SaemixData"): internal function to initialise object, not to be used

plot signature(x = "SaemixData"): plot the data
print signature(x = "SaemixData"): prints details about the object (more extensive than show)
read.saemixData signature(object = "SaemixData"): internal function, not to be used
showall signature(object = "SaemixData"): shows all the elements in the object
show signature(object = "SaemixData"): prints details about the object</pre>
```

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summary signature(object = "SaemixData"): summary of the data. Returns a list with a number of elements extracted from the dataset (N: the number of subjects; nobs: the total number of observations; nind.obs: a vector giving the number of observations for each subject; id: subject ID; x: predictors; y: response, and, if present in the data, covariates: the covariates (as many lines as observations) and ind.covariates: the individual covariates (one line per individual).

subset signature(object = "SaemixData"): extract part of the data; this function will operate
 on the rows of the dataset (it can be used for instance to extract the data corresponding to the
 first ten subjects)

#### Author(s)

Emmanuelle Comets <a href="mailto:emmanuelle.comets@inserm.fr">emmanuelle.comets@inserm.fr</a>, Audrey Lavenu, Marc Lavielle.

#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
saemixData,SaemixModel, saemixControl,saemix
```

```
showClass("SaemixData")
# Specifying column names
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
# Specifying column numbers
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=1,name.predictors=c(2,3),name.response=c(4), name.covariates=5:6,
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
# No column names specified, using automatic recognition of column names
data(PD1.saemix)
saemix.data<-saemixData(name.data=PD1.saemix,header=TRUE,</pre>
 name.covariates = c("gender"), units = list(x = "mg", y = "-", covariates = c("-")))
```

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saemixModel	Function to create a SaemixModel object	

# **Description**

This function creates a SaemixModel object. The two mandatory arguments are the name of a R function computing the model in the SAEMIX format (see details and examples) and a matrix psi0 giving the initial estimates of the fixed parameters in the model, with one row for the population mean parameters and one row for the covariate effects (see documentation).

# Usage

```
saemixModel(model, psi0, description = "", error.model = character(),
  transform.par = numeric(), fixed.estim = numeric(),
  covariate.model = matrix(nrow = 0, ncol = 0),
  covariance.model = matrix(nrow = 0, ncol = 0),
  omega.init = matrix(nrow = 0, ncol = 0), error.init = numeric(),
  name.modpar = character())
```

# **Arguments**

8	
model	name of the function used to compute the structural model. The function should return a vector of predicted values given a matrix of individual parameters, a vector of indices specifying which records belong to a given individual, and a matrix of dependent variables (see example below).
psi0	a matrix with a number of columns equal to the number of parameters in the model, and one (when no covariates are available) or two (when covariates enter the model) giving the initial estimates for the fixed effects. The column names of the matrix should be the names of the parameters in the model, and will be used in the plots and the summaries. When only the estimates of the mean parameters are given, psi0 may be a named vector.
descriptio	a character string, giving a brief description of the model or the analysis
error.mode	type of residual error model (valid types are constant, proportional, combined and exponential). Defaults to constant
transform.	par the distribution for each parameter (0=normal, 1=log-normal, 2=probit, 3=logit). Defaults to a vector of 1s (all parameters have a log-normal distribution)
fixed.esti	whether parameters should be estimated (1) or fixed to their initial estimate (0). Defaults to a vector of 1s
covariate.model	
	a matrix giving the covariate model. Defaults to no covariate in the model
covariance	e.model

a square matrix of size equal to the number of parameters in the model, giving the variance-covariance matrix of the model: 1s correspond to estimated variances (in the diagonal) or covariances (off-diagonal elements). Defaults to the identity matrix 48 saemixModel

omega.init	a square matrix of size equal to the number of parameters in the model, giving the initial estimate for the variance-covariance matrix of the model. Defaults to the identity matrix
error.init	a vector of size 2 giving the initial value of a and b in the error model. Defaults to 1 for each estimated parameter in the error model
name.modpar	names of the model parameters, if they are not given as the column names (or names) of psi0

# **Details**

This function is the user-friendly constructor for the SaemixModel object class.

# Value

A SaemixModel object (see saemixModel).

#### Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

# See Also

SaemixData,SaemixModel, saemixControl,saemix

```
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]</pre>
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
}
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3, byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
```

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SaemixModel-class

Class "SaemixModel"

# **Description**

An object of the SaemixModel class, representing a non-linear mixed-effect model structure, used by the SAEM algorithm.

# **Objects from the Class**

An object of the SaemixModel class can be created by using the function saemixModel

#### Slots

model: function used to compute the structural model. The function should return a vector of predicted values given a matrix of individual parameters, a vector of indices specifying which records belong to a given individual, and a matrix of dependent variables (see examples).

description: model description (optional) as a character string

psi0: a matrix with a number of columns equal to the number of parameters in the model, and one (when no covariates are available) or two (when covariates enter the model) giving the initial estimates for the fixed effects. The column names of the matrix should be the names of the parameters in the model, and will be used in the plots and the summaries

transform.par: the distribution for each parameter (0=normal, 1=log-normal, 2=probit, 3=logit). Defaults to a vector of 1s (all parameters have a log-normal distribution)

fixed.estim: whether parameters should be estimated (1) or fixed to their initial estimate (0). Defaults to a vector of 1s

error.model: name of the residual error model

covariate.model: a matrix giving the covariate model. Defaults to no covariate in the model (empty matrix)

betaest.model: a matrix giving the effects model (internal)

covariance.model: a square matrix of size equal to the number of parameters in the model, giving the variance-covariance matrix of the model: 1s correspond to estimated variances (in the diagonal) or covariances (off-diagonal elements). Defaults to the identity matrix

omega.init: a square matrix of size equal to the number of parameters in the model, giving the initial estimate for the variance-covariance matrix of the model. Defaults to the identity matrix

error.init: a vector of size 2 giving the initial value of a and b in the error model. Defaults to 1 for each estimated parameter in the error model

nb.parameters: number of parameters

name.modpar: names of the model parameters

name.fixed: names of the fixed effects estimated in the model

name.random: names of the random effects estimated in the model

name.res: names of the parameters of the residual error model

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```
name.predictors: names of the predictors (X)
name.X: name of the predictor used in graphs
name.response: name of the response (Y)
name.cov: name of the covariates
indx.fix: index of estimated fixed effects (internal)
indx.cov: index of estimated fixed effects associated with covariate effects (internal)
indx.omega: index of estimated random effects (internal)
indx.res: index of parameters of the residual error model (internal)
Mcovariates: matrix of the covariates (internal)
```

#### Methods

```
[<- signature(x = "SaemixModel"): replace elements of object
[ signature(x = "SaemixModel"): access elements of object
initialize signature(.Object = "SaemixModel"): internal function to initialise object, not to be used

plot signature(x = "SaemixModel"): plot results (see saemix.plot.data
print signature(x = "SaemixModel"): prints details about the object
showall signature(object = "SaemixModel"): prints an extensive summary of the object
show signature(object = "SaemixModel"): prints a short summary of the object
summary signature(object = "SaemixModel"): summary of the model</pre>
```

#### Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

## References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

# See Also

```
SaemixData, saemixModel, saemixControl, saemix
```

```
showClass("SaemixModel")
```

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SaemixObject-class Class "SaemixObject"

# **Description**

An object of the SaemixObject class, storing the results obtained by a call to the SAEM algorithm

#### **Details**

Details of the algorithm can be found in the pdf file accompanying the package.

# **Objects from the Class**

Objects are created by a call saemix().

#### **Slots**

data: Object of class "SaemixData" an object of the SaemixData class, representing a longitudinal data structure, used by the SAEM algorithm. See SaemixData

model: Object of class "SaemixModel" an object of the SaemixModel class, representing the structure a non-linear mixed effect model, used by the SAEM algorithm. See SaemixModel

results: Object of class "SaemixRes" ~~

rep.data: Object of class "SaemixRepData" (internal use only) the data part, replicated a number of times equal to the number of chains used in the SAEM algorithm (see documentation for details). Not intended to be accessed to directly by the user.

sim.data: Object of class "SaemixSimData" (internal use only) data simulated according to the design in data, with the model in model and the parameters estimated by the SAEM algorithm, after a call to simul.saemix (see documentation for details). Not intended to be accessed to directly by the user.

options: Object of class "list" a list of options containing variables controlling the algorithm prefs: Object of class "list" a list of graphical preferences applied to plots

```
[<- signature(x = "SaemixObject"): replace elements of object
[ signature(x = "SaemixObject"): access elements of object
initialize signature(.Object = "SaemixObject"): internal function to initialise object, not to
    be used

plot signature(x = "SaemixObject"): plot results (see saemix.plot.data

predict signature(object = "SaemixObject"): compute model predictions

print signature(x = "SaemixObject"): prints details about the object

showall signature(object = "SaemixObject"): prints an extensive summary of the object

show signature(object = "SaemixObject"): prints a short summary of the object</pre>
```

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```
summary signature(object = "SaemixObject"): summary of the results
```

coef signature(object = "SaemixObject"): extracts coefficients. Returns a list with components fixed (estimated fixed effects), population (population parameter estimates, including covariate effects: a list with two components map and cond), individual (individual parameter estimates: a list with two components map and cond). For population and individual, the map component of the list gives the MAP estimates (the mode of the distribution) while the cond component gives the conditional mean estimates. Some components may be missing (eg MAP estimates) if they have not been computed during or after the fit.

#### Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

#### References

Kuhn E, Lavielle M. Maximum likelihood estimation in nonlinear mixed effects models. Computational Statistics and Data Analysis 49, 4 (2005), 1020-1038.

Comets E, Lavenu A, Lavielle M. SAEMIX, an R version of the SAEM algorithm. 20th meeting of the Population Approach Group in Europe, Athens, Greece (2011), Abstr 2173.

#### See Also

```
SaemixData,SaemixModel, saemixControl,saemix,plot.saemix, saemix.plot.data
```

# **Examples**

```
showClass("SaemixObject")
```

show-methods

Methods for Function show

# **Description**

Prints a short summary of an object

```
signature(x = "ANY") Default show function
signature(x = "SaemixData") Prints a short summary of a SaemixData object
signature(x = "SaemixModel") Prints a short summary of a SaemixModel object
signature(x = "SaemixObject") Prints a short summary of the results from a SAEMIX fit
signature(x = "SaemixRes") Not user-level
signature(object = "SaemixRepData") Prints a short summary of a SaemixRepData object
signature(object = "SaemixSimData") Prints a short summary of a SaemixSimData object
```

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showall

Prints out an extensive summary of an object

# **Description**

This function shows an extensive summary of an object, and is used mainly to visualise the majority of the elements of an object

# Usage

```
showall(object)
```

# **Arguments**

object

showall methods are available for objects of typeSaemixData, SaemixModel and SaemixObject

# **Details**

None

# Value

None

# See Also

SaemixData,SaemixModel, SaemixObject

```
# A SaemixData object
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response = c("Concentration"), name.covariates = c("Weight", "Sex"),\\
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
showall(saemix.data)
# A SaemixModel object
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]</pre>
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
```

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```
}
saemix.model<-saemixModel(model=model1cpt,
    description="One-compartment model with first-order absorption",
    psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3, byrow=TRUE,
    dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
    covariate.model=matrix(c(0,1,0,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
    covariance.model=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),
    omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="constant")
showall(saemix.model)
</pre>
```

showall-methods

Methods for Function showall

# **Description**

Prints an extensive summary of an object

## Methods

```
signature(x = "SaemixData") Prints a extensive summary of a SaemixData object
signature(x = "SaemixModel") Prints a extensive summary of a SaemixModel object
signature(x = "SaemixObject") Prints a extensive summary of the results from a SAEMIX fit
signature(x = "SaemixRes") Not user-level
```

simul.saemix

Perform simulations under the model

# **Description**

This function is used to simulate from the model. It can be called with the estimated parameters (the default), the initial parameters, or with a set of parameters. The original design can be used in the simulations, or a different dataset may be used with the same structure (covariates) as the original design. This function is not yet implemented.

# Usage

```
simul.saemix(saemixObject, nsim = saemixObject["options"]$nb.sim,
    predictions = TRUE, res.var = TRUE, uncertainty = FALSE)
```

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## **Arguments**

saemixObject an object returned by the saemix function

nsim Number of simulations to perform. Defaults to the nb.simpred element in options

predictions Whether the simulated parameters should be used to compute predictions. Defaults to TRUE

res.var Whether residual variability should be added to the predictions. Defaults to TRUE

# Details

uncertainty

This function is used to produce Visual Predictive Check graphs, as well as to compute the normalised prediction distribution errors (npde).

Uses uncertainty (currently not implemented). Defaults to FALSE

# Author(s)

Emmanuelle Comets <emmanuelle.comets@inserm.fr>, Audrey Lavenu, Marc Lavielle.

#### References

Brendel, K, Comets, E, Laffont, C, Laveille, C, Mentre, F. Metrics for external model evaluation with an application to the population pharmacokinetics of gliclazide, Pharmaceutical Research 23 (2006), 2036-2049.

Holford, N. The Visual Predictive Check: superiority to standard diagnostic (Rorschach) plots (Abstract 738), in: 14th Meeting of the Population Approach Group in Europe, Pamplona, Spain, 2005.

#### See Also

SaemixObject, saemix.plot.data, saemix.plot.convergence, saemix.plot.llis, saemix.plot.randeff, saemix.plot.obsvspred, saemix.plot.fits, saemix.plot.parcov, saemix.plot.distpsi, saemix.plot.scatterressaemix.plot.vpc

nods for subset	subset-methods	
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# **Description**

This method is used to select a subset of an SaemixData object

```
signature(x = "ANY") Default subset function
signature(x = "SaemixData") A subset of the dataset can be selected (see subset )
```

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summary-methods

Methods for Function summary

# Description

Methods for function summary

#### Methods

```
signature(x = "ANY") default summary function ?
signature(x = "SaemixData") summary of the data
signature(x = "SaemixModel") summary of the model
signature(x = "SaemixObject") summary of an SaemixObject
```

testnpde

Tests for normalised prediction distribution errors

# **Description**

Performs tests for the normalised prediction distribution errors returned by npde

# Usage

```
testnpde(npde)
```

# **Arguments**

npde

the vector of prediction distribution errors

# **Details**

Given a vector of normalised prediction distribution errors (npde), this function compares the npde to the standardised normal distribution N(0,1) using a Wilcoxon test of the mean, a Fisher test of the variance, and a Shapiro-Wilks test for normality. A global test is also reported.

The helper functions kurtosis and skewness are called to compute the kurtosis and skewness of the distribution of the npde.

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#### Value

```
a list containing 4 components:
```

Wilcoxon test of mean=0

compares the mean of the npde to 0 using a Wilcoxon test

variance test compares the variance of the npde to 1 using a Fisher test

SW test of normality

compares the npde to the normal distribution using a Shapiro-Wilks test

global test

an adjusted p-value corresponding to the minimum of the 3 previous p-values multiplied by the number of tests (3), or 1 if this p-value is larger than 1.

# Author(s)

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#### References

K. Brendel, E. Comets, C. Laffont, C. Laveille, and F. Mentr\'e. Metrics for external model evaluation with an application to the population pharmacokinetics of gliclazide. *Pharmaceutical Research*, 23:2036–49, 2006.

# See Also

```
saemix, saemix.plot.npde
```

theo.saemix

Pharmacokinetics of theophylline, in SAEM format

# **Description**

The theo.saemix data frame has 132 rows and 6 columns of data from an experiment on the pharmacokinetics of theophylline. A column with gender was added to the original data for demo purposes, and contains simulated data.

#### **Usage**

theo.saemix

# Format

This data frame contains the following columns:

**Id:** an ordered factor with levels 1, ..., 12 identifying the subject on whom the observation was made. The ordering is by Time at which the observation was made.

**Dose:** dose of theophylline administered orally to the subject (mg/kg).

**Time:** time since drug administration when the sample was drawn (hr).

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**Concentration:** the ophylline concentration in the sample (mg/L).

Weight: weight of the subject (kg).

**Sex:** gender of the subject (0=men, 1=women).

# **Details**

Boeckmann, Sheiner and Beal (1994) report data from a study by Dr. Robert Upton of the kinetics of the anti-asthmatic drug theophylline. Twelve subjects were given oral doses of theophylline then serum concentrations were measured at 11 time points over the next 25 hours. In the present package *npde*, we removed the data at time 0.

These data are analyzed in Davidian and Giltinan (1995) and Pinheiro and Bates (2000) using a two-compartment open pharmacokinetic model.

These data are also available in the library datasets under the name Theoph in a slightly modified format and including the data at time 0.

#### **Source**

Boeckmann, A. J., Sheiner, L. B. and Beal, S. L. (1994), *NONMEM Users Guide: Part V*, NON-MEM Project Group, University of California, San Francisco.

Davidian, M. and Giltinan, D. M. (1995) *Nonlinear Models for Repeated Measurement Data*, Chapman & Hall (section 5.5, p. 145 and section 6.6, p. 176)

Pinheiro, J. C. and Bates, D. M. (2000) *Mixed-effects Models in S and S-PLUS*, Springer (Appendix A.29)

```
data(theo.saemix)
saemix.data<-saemixData(name.data=theo.saemix,header=TRUE,sep=" ",na=NA,</pre>
 name.group=c("Id"),name.predictors=c("Dose","Time"),
 name.response=c("Concentration"), name.covariates=c("Weight", "Sex"),
 units=list(x="hr",y="mg/L",covariates=c("kg","-")), name.X="Time")
model1cpt<-function(psi,id,xidep) {</pre>
 dose<-xidep[,1]
 tim<-xidep[,2]
 ka<-psi[id,1]
 V<-psi[id,2]
 CL<-psi[id,3]
 k<-CL/V
 ypred<-dose*ka/(V*(ka-k))*(exp(-k*tim)-exp(-ka*tim))</pre>
 return(ypred)
# Default model, no covariate
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3,byrow=TRUE,
 \label{eq:continuous} dimnames=list(NULL, \ c("ka","V","CL"))), transform.par=c(1,1,1))
```

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```
# Note: remove the options save=FALSE and save.graphs=FALSE
# to save the results and graphs
saemix.options<-list(seed=632545,save=FALSE,save.graphs=FALSE)</pre>
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
# Model with covariates
saemix.model<-saemixModel(model=model1cpt,</pre>
 description="One-compartment model with first-order absorption",
 psi0=matrix(c(1.,20,0.5,0.1,0,-0.01),ncol=3,byrow=TRUE,
 dimnames=list(NULL, c("ka","V","CL"))),transform.par=c(1,1,1),
 covariate.model=matrix(c(0,0,1,0,0,0),ncol=3,byrow=TRUE),fixed.estim=c(1,1,1),
 covariance.model=matrix(c(1,0,0,0,1,1,0,1,1),ncol=3,byrow=TRUE),
 omega.init=matrix(c(1,0,0,0,1,0,0,0,1),ncol=3,byrow=TRUE),error.model="combined")\\
# saemix.options<-list(seed=39546,save=FALSE,save.graphs=FALSE)
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
```

yield.saemix

Wheat yield in crops treated with fertiliser, in SAEM format

# **Description**

yield.saemix contains data from winter wheat experiments.

# Usage

```
yield.saemix
```

#### **Format**

This data frame contains the following columns:

site: the site number

dose: dose of nitrogen fertiliser (kg/ha).

yield: grain yield (kg/ha).

**soil.nitrogen:** end-of-winter mineral soil nitrogen (NO3- plus NH4+) in the 0 to 90 cm layer was measured on each site/year (kg/ha).

# **Details**

The data in the yield.saemix comes from 37 winter wheat experiments carried out between 1990 and 1996 on commercial farms near Paris, France. Each experiment was from a different site. Two soil types were represented, a loam soil and a chalky soil. Common winter wheat varieties were used. Each experiment consisted of five to eight different nitrogen fertiliser rates, for a total of

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224 nitrogen treatments. Nitrogen fertilizer was applied in two applications during the growing season. For each nitrogen treatment, grain yield (adjusted to 150 g.kg-1 grain moisture content) was measured. In addition, end-of-winter mineral soil nitrogen (NO3- plus NH4+) in the 0 to 90 cm layer was measured on each site-year during February when the crops were tillering. Yield and end-of-winter mineral soil nitrogen measurements were in the ranges 3.44-11.54 t.ha-1, and 40-180 kg.ha-1 respectively.

#### Source

Makowski, D., Wallach, D., and Meynard, J.-M (1999). Models of yield, grain protein, and residual mineral nitrogen responses to applied nitrogen for winter wheat. Agronomy Journal 91: 377-385.

```
data(yield.saemix)
saemix.data<-saemixData(name.data=yield.saemix,header=TRUE,name.group=c("site"),</pre>
 name.predictors=c("dose"),name.response=c("yield"),
 name.covariates=c("soil.nitrogen"),units=list(x="kg/ha",y="t/ha",
 covariates=c("kg/ha")))
# Model: linear + plateau
yield.LP<-function(psi,id,xidep) {</pre>
# input:
   psi : matrix of parameters (3 columns, ymax, xmax, slope)
   id : vector of indices
   xidep: dependent variables (same nb of rows as length of id)
   a vector of predictions of length equal to length of id
 x < -xidep[,1]
 ymax<-psi[id,1]</pre>
 xmax<-psi[id,2]
 slope<-psi[id,3]</pre>
  f<-ymax+slope*(x-xmax)
# cat(length(f)," ",length(ymax),"\n")
 f[x>xmax]<-ymax[x>xmax]
 return(f)
}
saemix.model<-saemixModel(model=yield.LP,description="Linear plus plateau model",</pre>
 psi0=matrix(c(8,100,0.2,0,0),ncol=3,byrow=TRUE,dimnames=list(NULL,
 c("Ymax", "Xmax", "slope"))),covariate.model=matrix(c(0,0,0),ncol=3,byrow=TRUE),
 transform.par=c(0,0,0), covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3,
 byrow=TRUE),error.model="constant")
saemix.options<-list(algorithms=c(1,1,1),nb.chains=1,seed=666,</pre>
   save=FALSE, save.graphs=FALSE)
# Plotting the data
plot(saemix.data,xlab="Fertiliser dose (kg/ha)", ylab="Wheat yield (t/ha)")
# Not run (strict time constraints for CRAN)
# saemix.fit<-saemix(saemix.model,saemix.data,saemix.options)</pre>
```

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```
# Comparing the likelihoods obtained by linearisation and importance sampling
# to the likelihood obtained by Gaussian Quadrature
# Not run
# saemix.fit<-llgq.saemix(saemix.fit)</pre>
# cat("LL by Importance sampling, LL_IS=",saemix.fit["results"]["ll.is"],"\n")
# cat("LL by linearisation, LL_lin=",saemix.fit["results"]["ll.lin"],"\n")
# cat("LL by Gaussian Quadrature, LL_GQ=",saemix.fit["results"]["ll.gq"],"\n")
# Testing for an effect of covariate soil.nitrogen on Xmax
saemix.model2<-saemixModel(model=yield.LP,description="Linear plus plateau model",</pre>
 psi0=matrix(c(8,100,0.2,0,0,0),ncol=3,byrow=TRUE,dimnames=list(NULL,
 c("Ymax","Xmax","slope"))),covariate.model=matrix(c(0,1,0),ncol=3,byrow=TRUE),
 transform.par=c(0,0,0), covariance.model=matrix(c(1,0,0,0,1,0,0,0,1), ncol=3,
 byrow=TRUE),error.model="constant")
# saemix.fit2<-saemix(saemix.model2,saemix.data,saemix.options)</pre>
# BIC for the two models
# cat("Model without covariate, BIC=",saemix.fit["results"]["bic.is"],"\n")
# cat("Model with covariate, BIC=",saemix.fit2["results"]["bic.is"],"\n")
# pval<-1-pchisq(-2*saemix.fit["results"]["11.is"]+2*saemix.fit2["results"]["11.is"],1)</pre>
# cat("
               LRT: p=",pval,"\n")
}
```

[-methods

Methods for "get" Function [

# Description

Methods for function [, to allow access to the value of an element within an S4 object

```
signature(x = "SaemixData") access elements of an object SaemixData
signature(x = "SaemixModel") access elements of an object SaemixModel
signature(x = "SaemixObject") access elements of an object SaemixObject
signature(x = "SaemixRepData") access elements of an object SaemixRepData
signature(x = "SaemixRes") access elements of an object SaemixRes
signature(x = "SaemixSimData") access elements of an object SaemixSimData
```

[<-methods

[<--methods

Methods for "set" Function [<-

# Description

Methods for function [<-, to allow replacements of the value of an element within an S4 object

```
signature(x = "SaemixData") replace elements of an object SaemixData
signature(x = "SaemixModel") replace elements of an object SaemixModel
signature(x = "SaemixObject") replace elements of an object SaemixObject
signature(x = "SaemixRepData") replace elements of an object SaemixRepData
signature(x = "SaemixRes") replace elements of an object SaemixRes
signature(x = "SaemixSimData") replace elements of an object SaemixSimData
```

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