# Package 'spm'

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Type Package	
Title Stochastic Process M	Iodeling (SPM)
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<b>Description</b> Stochastic Pro	ocess Modeling
License GPL	
<b>Imports</b> Rcpp (>= 0.11.1)	, RcppArmadillo (>= 0.4.200.0)
LinkingTo Rcpp, RcppAr	madillo
<b>Depends</b> deSolve,mice,sas	s7bdat,RcppArmadillo
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# Description

Stochastic Process Modeling (SPM) perform estimation of parameters of some stochastic process of interest. This package is mainly for estimating parameters from clinical traits.

# **Details**

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Package: spm Type: Package Version: 1.0 Date: 2015-09-05 License: GPL

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#### References

Anatoli I. Yashin, Konstantin G. Arbeev, Igor Akushevich, Aliaksandr Kulminski, Lucy Akushevich, Svetlana V. Ukraintseva, Stochastic model for analysis of longitudinal data on aging and mortality, Mathematical Biosciences, Volume 208, Issue 2, August 2007, Pages 538-551, ISSN 0025-5564, http://dx.doi.org/10.1016/j.mbs.2006.11.006.

Anatoli I. Yashin, Konstantin G. Arbeev, Aliaksandr Kulminski, Igor Akushevich, Lucy Akushevich, Svetlana V. Ukraintseva, Health decline, aging and mortality: how are they related? Biogerontology June 2007, Volume 8, Issue 3, pp 291-302

#### **Examples**

```
library(spm)
# Reading longitude data:
longdat <- read.csv(system.file("data","longdat.csv",package="spm"))
# Prepare data for optimization:
vitstat <- read.csv(system.file("data","vitstat.csv",package="spm"))
# Remove unneeded NAs:
longdat.nonan <- longdat[which(is.na(longdat$Age) == F),]
vitstat.nonan <- vitstat[which(is.na(vitstat$BirthCohort) == F),]
data=prepare_data(longdat=longdat.nonan, vitstat=vitstat.nonan,interval=1, col.status="IsDead", col.id="ID"
# Parameters estimation:
pars=spm(data,k = 1)
pars</pre>
```

prepare\_data

Output values include: 1). Database, prepared for (slow) continuous optimization (with integral). 2). Database, prepared for (quick) discrete optimization (which is used for parameter estimations)

## Description

Output values include: 1). Database, prepared for (slow) continuous optimization (with integral). 2). Database, prepared for (quick) discrete optimization (which is used for parameter estimations)

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#### **Usage**

```
prepare_data(longdat, vitstat, interval = 1, col.status = "IsDead",
  col.id = "ID", col.age = "Age", col.age.event = "LSmort",
  covariates = c("DBP", "BMI", "DBP1", "DBP2", "Weight", "Height"),
  verbose = T)
```

#### **Arguments**

longdat A table with longitude records. vitstat A table with vital statistics. col.status A name of column containing status variable (0/1 which indicate alive/dead). col.id A name of column containing patient ID. A name of age column. col.age - A name of event column. col.age.event covariates A list of covariates. A verbosing output indicator, default TRUE. verbose

#### Value

A list of two elements: first element contains a data table for continuous optimization and second element contains a data table for quick discrete optimization used in estimation of starting point.

## **Examples**

```
library(spm)
#Reading longitude data:
longdat <- read.csv(system.file("data","longdat.csv",package="spm"))
# Prepare data for optimization:
vitstat <- read.csv(system.file("data","vitstat.csv",package="spm"))
# Remove unneeded NAs:
longdat.nonan <- longdat[which(is.na(longdat$Age) == F),]
vitstat.nonan <- vitstat[which(is.na(vitstat$BirthCohort) == F),]
data=prepare_data(longdat=longdat.nonan, vitstat=vitstat.nonan,interval=1, col.status="IsDead", col.id="ID"
# Parameters estimation:
pars=spm(data,k = 1)
pars</pre>
```

sim

Multi-dimension simulation function It uses a, f1, Q, f, b, mu0 and theta as input parameters.

## Description

Multi-dimension simulation function It uses a, f1, Q, f, b, mu0 and theta as input parameters.

#### Usage

```
sim(N = 100, a = -0.05, f1 = 80, Q = 2e-08, f = 80, b = 5, mu0 = 1e-05, theta = 0.08, ystart = c(80), tstart = 30, tend = 105, dt = 1, k = 1)
```

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## **Arguments**

N	Number of individuals
а	A k by k matrix, which characterize the rate of the adaptive response.
f1	A particular state, which if a deviation from the normal (or optimal). This is a vector with length of $\mathbf{k}$ .
Q	A matrix k by k, which is a non-negative-definite symmetric matrix.
f	A vector-function (with length k) of the normal (or optimal) state.
b	A diffusion coefficient, k by k matrix.
mu0	mortality at start period of time.
theta	A displacement coefficient of the Gompertz function.
ystart	A vector with length equal to number of dimensions used, defines starting values of covariates.
tstart	A number that defines starting time (30 by default).
tend	A number, defines final time (105 by default).
dt	A time step (1 by default).
k	number of dimensions $(k = 1 \text{ by default})$ .

## Value

A table with simulated data.

# **Examples**

```
library(spm)
data <- sim(N=1000, ystart=c(75, 94), k=1)
head(data)</pre>
```

simdata

Function that simulates data using u, R, epsilon, mu0, b, Q, theta

## Description

Function that simulates data using u, R, epsilon, mu0, b, Q, theta

# Usage

```
simdata(N = 100, u = 8, R = 0.95, epsilon = 5, mu0 = 2e-05, b = 10, Q = 2e-08, theta = 0.08, tstart = 30, ystart = 80, dt = 1, tmax = 105, k = 1)
```

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## **Arguments**

N	Number of individuals
u	A drift vector with length of k.
R	A k by k regression matrix.
epsilon	A time-dependent normally distributed random vector (size=k).
mu0	mortality at start period of time.
b	A diffusion coefficient, k by k matrix.
Q	A matrix k by k, which is a non-negative-definite symmetric matrix.
theta	A displacement coefficient of the Gompertz function.
tstart	A number that defines starting time (30 by default).
ystart	A vector with length equal to number of dimensions used, defines starting values of covariates.
dt	A time step (1 by default).
k	Number of dimensions $(k = 1 \text{ by default})$ .

A number, defines final time (105 by default).

## Value

tend

A table with simulated data.

## **Examples**

```
library(spm)
data <- simdata(N=1000, ystart=c(75, 94), k=1)
head(data)</pre>
```

simdata\_cont

Simulation function for continuous trait.

# Description

Simulation function for continuous trait.

## Usage

```
simdata\_cont(N = 10, aH = -0.05, f1H = 80, QH = 2e-07, fH = 80, bH = 5, mu0H = 2e-05, thetaH = 0.08, step = 0.05, tstart = 30, tend = 105, ystart = 80, sd0 = 4)
```

# Arguments

N	Number of individuals.
tstart	A number that defines starting time (30 by default).
tend	A number, defines final time (105 by default).
ystart	A vector with length equal to number of dimensions used, defines starting values of covariates.

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a	A k by k matrix, which characterize the rate of the adaptive response.
f1	A particular state, which if a deviation from the normal (or optimal). This is a vector with length of ${\bf k}$ .
Q	A matrix k by k, which is a non-negative-definite symmetric matrix.
f	A vector-function (with length k) of the normal (or optimal) state.
b	A diffusion coefficient, k by k matrix.
mu0	mortality at start period of time.
theta	A displacement coefficient of the Gompertz function.
k	number of dimensions $(k = 1 \text{ by default})$ .

#### Value

A table with simulated data.

# **Examples**

```
library(spm)
dat <- simdata_cont(N=2500)
dat</pre>
```

spm

Stochastic Process Modelling (SPM) A main function that estimates parameters a, f1, Q, f, b, mu0, theta from given dataset.

# Description

Stochastic Process Modelling (SPM) A main function that estimates parameters  $a,\,f1,\,Q,\,f,\,b,\,mu0,$  theta from given dataset.

# Usage

```
spm(dat, k = 2, verbose = F, tol = NULL)
```

## **Arguments**

dat A dataset.

k Number of dimensions.

verbose A verbosing output indicator.

tol A tolerance threshold for matrix inversion.

## Value

A list of (1) Estimated starting point (from quick discrete optimization) and (2) Estimated coefficients.

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#### **Examples**

```
library(spm)
# Reading longitude data:
longdat <- read.csv(system.file("data","longdat.csv",package="spm"))
# Prepare data for optimization:
vitstat <- read.csv(system.file("data","vitstat.csv",package="spm"))
# Remove unneeded NAs:
longdat.nonan <- longdat[which(is.na(longdat$Age) == F),]
vitstat.nonan <- vitstat[which(is.na(vitstat$BirthCohort) == F),]
data=prepare_data(longdat=longdat.nonan, vitstat=vitstat.nonan,interval=1, col.status="IsDead", col.id="ID"
# Parameters estimation:
pars=spm(data,k = 1)
pars</pre>
```

spm\_integral\_MD

Continuous multi-dimensional optimization It is much slower that discrete but more precise and can handle time intervals with different lengths.

#### **Description**

Continuous multi-dimensional optimization It is much slower that discrete but more precise and can handle time intervals with different lengths.

## Usage

```
spm_integral_MD(dat, parameters, k, verbose = F)
```

## **Arguments**

dat A data table.

parameters A starting pont (a vector). k A number of dimensions.

verbose An indicator of verbosing output.

tol A tolerance threshold for matrix inversion.

#### Value

A list of two elements: (1) parameters a, f1, Q, f, b, mu0, theta; (2) An output from "optim" function used for maximum likelihood estimation.

## **Examples**

```
#'library(spm)
# Reading longitude data:
longdat <- read.csv(system.file("data","longdat.csv",package="spm"))
# Prepare data for optimization:
vitstat <- read.csv(system.file("data","vitstat.csv",package="spm"))
# Remove unneeded NAs:
longdat.nonan <- longdat[which(is.na(longdat$Age) == F),]
vitstat.nonan <- vitstat[which(is.na(vitstat$BirthCohort) == F),]
dat=prepare_data(longdat=longdat.nonan, vitstat=vitstat.nonan,interval=1, col.status="IsDead", col.id="ID",</pre>
```

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```
# Parameters estimation: dat<-[,1:6] pars=spm_integral_MD(dat, parameters=c(-0.05, 80, 2e-8, 80, 5, 2e-5, 0.08), k=1) pars
```

spm\_quick\_MD

Discrete multi-dimensional optimization It is way much faster that continuous (but less precise) and used mainly in estimation of starting point.

#### **Description**

Discrete multi-dimensional optimization It is way much faster that continuous (but less precise) and used mainly in estimation of starting point.

#### Usage

```
spm_quick_MD(dat, k = 2, theta_range = seq(0.078, 0.082, by = 1e-04), tol = NULL)
```

## **Arguments**

dat A data table.

k A number of dimensions.

theta\_range A range of theta parameter (axe displacement of Gompertz function).

tol A tolerance threshold for matrix inversion.

#### Value

A list of two elements: (1) parameters u, R, b, epsilon, Q, mu0, theta and (2) parameters a, f1, Q, f, b, mu0, theta. Note: b and mu0 from first list are different from b and mu0 from the second list.

#### **Examples**

```
#'library(spm)
# Reading longitude data:
longdat <- read.csv(system.file("data","longdat.csv",package="spm"))
# Prepare data for optimization:
vitstat <- read.csv(system.file("data","vitstat.csv",package="spm"))
# Remove unneeded NAs:
longdat.nonan <- longdat[which(is.na(longdat$Age) == F),]
vitstat.nonan <- vitstat[which(is.na(vitstat$BirthCohort) == F),]
data=prepare_data(longdat=longdat.nonan, vitstat=vitstat.nonan,interval=1, col.status="IsDead", col.id="ID"
# Parameters estimation:
pars=spm_quick_MD(data,k = 1)
pars</pre>
```

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