# Package 'SmoothHazard'

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Title Estimation of Smooth Hazard Models for Interval-Censored Data

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Description Estimation of two-state (survival) models and irreversible illness-death models with possibly interval-censored, left-truncated and right-censored data. Proportional intensities regression models can be specified to allow for covariates effects separately for each transition. We use either a parametric approach with Weibull baseline intensities or a semi-parametric approach with M-splines approximation of baseline intensities in order to obtain smooth estimates of the hazard functions. Parameter estimates are obtained by maximum likelihood in the parametric approach and by penalized maximum likelihood in the semi-parametric approach.

**Encoding** UTF-8

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**Imports** lava (>= 1.4.1), mvtnorm (>= 1.0-3)

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idm

Fit an illness-death model

#### **Description**

Fit an illness-death model using either a semi-parametric approach (penalized likelihood with an approximation of the transition intensity functions by linear combination of M-splines) or a parametric approach (specifying Weibull distributions on the transition intensities). Left-truncated, right-censored, and interval-censored data are allowed. State 0 corresponds to the initial state, state 1 to the transient one, state 2 to the absorbant one. The allowed transitions are:  $0 \rightarrow 1$ ,  $0 \rightarrow 2$  and  $1 \rightarrow 2$ .

## Usage

```
idm(
  formula01,
  formula02,
  formula12,
  data,
  maxiter = 200,
  eps = c(5, 5, 3),
  n.knots = c(7, 7, 7),
  knots = "equidistant",
  CV = FALSE,
  kappa = c(1000000, 500000, 20000),
  method = "Weib",
```

```
conf.int = 0.95,
print.iter = FALSE,
subset = NULL,
na.action = na.fail
)
```

#### **Arguments**

formula01

A formula specifying a regression model for the  $\emptyset$  --> 1 transition from the initial state to the transient state of the illness-death model. The right hand side of the formula specifies the covariate terms, and the left hand side must be an event history object as returned by the function Hist.

formula02

A formula specifying a regression model for the  $\emptyset$  --> 2 transition from the initial state to the absorbing state. The left hand side must be equal to the left hand side of formula01. If missing it is set to formula01.

formula12

A formula specifying a regression model for the 1 --> 2 transition from the transient state to the absorbing state. operator is not required. If missing it is set to formula01.

data

A data frame in which to interpret the variables of formula01, formula02 and formula12.

maxiter

Maximum number of iterations. The default is 200.

eps

A vector of 3 integers >0 used to define the power of three convergence criteria: 1. for the regression parameters, 2. for the likelihood, 3. for the second derivatives. The default is c(5,5,3) which is translated into convergence if the respective values change less then  $10^{-5}$  (for regression parameters and likelihood) and  $10^{-3}$  for the second derivatives between two iterations.

n.knots

For method="Splines" only, a vector of length 3 specifing the number of knots, one for each transition, for the M-splines estimate of the baseline intensities in the order  $\emptyset$  --> 1,  $\emptyset$  --> 2, 1 --> 2. The default is c(7,7,7). When knots are specified as a list this argument is ignored. The algorithm needs least 5 knots and at most 20 knots.

knots

Argument only active for the penalized likelihood approach method="Splines". There are three ways to control the placement of the knots between the smallest and the largest of all time points:

knots="equidistant" Knots are placed with same distance on the time scale.
knots="quantiles" Knots are placed such that the number of observations is
roughly the same between knots.

**knots=list()** List of 1 or 2 or three vectors. The list elements are the actual placements (timepoints) of the knots for the M-spline. The list may contain one vector of placements for each transition in the order  $\emptyset$  --> 1,  $\emptyset$  --> 2, 1 --> 2. If only vector is specifified the knots are used for all transitions. If only 2 vectors are specifified, the knots for the  $\emptyset$  --> 1 transition are also used for the 1 --> 2 transition.

The algorithm needs at least 5 knots and allows no more than 20 knots.

CV Binary variable equals to 1 when search (by approximated cross validation) of the smoothing parameters kappa and 0 otherwise. Argument for the penalized

likelihood approach. The default is 0.

kappa Argument only active for the penalized likelihood approach method="Splines".

A vector with 3 positive values (smoothing parameters), one for each transition, in the order  $0 \rightarrow 1$ ,  $0 \rightarrow 2$  and  $1 \rightarrow 2$ .. If CV=1 these are used as starting values

for a cross validation search to optimize kappa.

method type of estimation method: "Splines" for a penalized likelihood approach with

approximation of the transition intensities by M-splines, "Weib" for a parametric approach with a Weibull distribution on the transition intensities. Default is

"Weib".

conf.int Level of confidence pointwise confidence intervals of the transition intensities,

i.e., a value between 0 and 1, the default is 0.95. The default is also used when conf.int=TRUE. To avoid computation of confidence intervals, set conf.int to

FALSE or NULL.

print.iter boolean parameter. Equals to TRUE to print the likelihood during the iteration

process, FALSE otherwise. Default is FALSE. This option is not running on Win-

dows.

subset expression indicating the subset of the rows of data to be used in the fit. All

observations are included by default.

na.action how NAs are treated. The default is first, any na.action attribute of data, second

a na.action setting of options, and third 'na.fail' if that is unset. The 'factory-

fresh' default is na.omit. Another possible value is NULL.

#### **Details**

The estimated parameters are obtained using the robust Marquardt algorithm (Marquardt, 1963) which is a combination between a Newton-Raphson algorithm and a steepest descent algorithm.

## Value

call the call that produced the result.

coef regression parameters.

loglik vector containing the log-likelihood without and with covariate.

cv vector containing the convergence criteria.

niter number of iterations.

converged integer equal to 1 when the model converged, 2, 3 or 4 otherwise.

modelPar Weibull parameters.

N number of subjects.

events 1 number of events  $0 \rightarrow 1$ .

events2 number of events  $0 \rightarrow 2$  or  $0 \rightarrow 1 \rightarrow 2$ .

NC vector containing the number of covariates on transitions  $0 \rightarrow 1$ ,  $0 \rightarrow 2$ ,  $1 \rightarrow 2$ .

responseTrans model response for the  $0 \rightarrow 1$  transition. Hist or Surv object.

responseAbs model response for the 0 -> 2 transition. Hist or Surv object.

time times for which transition intensities have been evaluated for plotting. Vector

in the Weibull approach. Matrix in the penalized likelihhod approach for which

the colums corresponds to the transitions  $0 \rightarrow 1$ ,  $1 \rightarrow 2$ ,  $0 \rightarrow 2$ .

intensity01 matched values of the intensities for transition  $0 \rightarrow 1$ .

lowerIntensity01

lower confidence intervals for the values of the intensities for transition  $0 \rightarrow 1$ .

upperIntensity01

upper confidence intervals for the values of the intensities for transition  $0 \rightarrow 1$ .

intensity02 matched values of the intensities for transition  $0 \rightarrow 2$ .

lowerIntensity02

lower confidence intervals for the values of the intensities for transition  $0 \rightarrow 2$ . upperIntensity02

upper confidence intervals for the values of the intensities for transition  $0 \rightarrow 2$ .

intensity 12 matched values of the intensities for transition  $1 \rightarrow 2$ .

lowerIntensity12

lower confidence intervals for the values of the intensities for transition  $1 \rightarrow 2$ .

upperIntensity12

upper confidence intervals for the values of the intensities for transition  $1 \rightarrow 2$ .

RR vector of relative risks.

V variance-covariance matrix derived from the Hessian of the log-likelihood if

using method="Weib" or, from the Hessian of the penalized log-likelihood if

using method="Splines".

se standart errors of the regression parameters.

Xnames01 names of covariates on  $0 \rightarrow 1$ . Xnames02 names of covariates on  $0 \rightarrow 2$ . Xnames12 names of covariates on  $1 \rightarrow 2$ .

knots01 knots to approximate by M-splines the intensity of the  $0 \rightarrow 1$  transition. knots02 knots to approximate by M-splines the intensity of the  $0 \rightarrow 2$  transition. knots12 knots to approximate by M-splines the intensity of the  $1 \rightarrow 2$  transition.

nknots01 number of knots on transition  $0 \rightarrow 1$ . nknots02 number of knots on transition  $0 \rightarrow 2$ . nknots12 number of knots on transition  $1 \rightarrow 2$ .

theta01 square root of splines coefficients for transition  $0 \rightarrow 1$ . theta02 square root of splines coefficients for transition  $0 \rightarrow 2$ . theta12 square root of splines coefficients for transition  $1 \rightarrow 2$ .

CV a binary variable equals to 1 when search of the smoothing parameters kappa by

approximated cross-validation, 1 otherwise. The default is 0.

kappa vector containing the smoothing parameters for transition  $0 \rightarrow 1$ ,  $0 \rightarrow 2$ ,  $1 \rightarrow 2$ 

used to estimate the model by the penalized likelihood approach.

CVcrit cross validation criteria.

DoF degrees of freedom of the model.

na.action observations deleted if missing values.

#### Author(s)

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#### References

D. Marquardt (1963). An algorithm for least-squares estimation of nonlinear parameters. *SIAM Journal of Applied Mathematics*, 431-441.

#### See Also

```
print.idm summary.idm predict.idm
```

```
library(lava)
library(prodlim)
set.seed(17)
d <- simulateIDM(100)</pre>
# right censored data
fitRC <- idm(formula01=Hist(time=observed.illtime,event=seen.ill)~X1+X2,</pre>
             formula02=Hist(time=observed.lifetime,event=seen.exit)~X1+X2,
             formula12=Hist(time=observed.lifetime,event=seen.exit)~X1+X2,data=d,
             conf.int=FALSE)
fitRC
set.seed(17)
d <- simulateIDM(300)</pre>
fitRC.splines <- idm(formula01=Hist(time=observed.illtime,event=seen.ill)~X1+X2,</pre>
             formula02=Hist(time=observed.lifetime,event=seen.exit)~X1+X2,
             formula12=Hist(time=observed.lifetime,event=seen.exit)~1,data=d,
             conf.int=FALSE,method="splines")
# interval censored data
fitIC <- idm(formula01=Hist(time=list(L,R),event=seen.ill)~X1+X2,</pre>
             formula02=Hist(time=observed.lifetime,event=seen.exit)~X1+X2,
             formula12=Hist(time=observed.lifetime,event=seen.exit)~X1+X2,data=d,
             conf.int=FALSE)
fitIC
    data(Paq1000)
    # Illness-death model with certif on the 3 transitions
    # Weibull parametrization and likelihood maximization
    fit.weib <- idm(formula02=Hist(time=t,event=death,entry=e)~certif,</pre>
                     formula01=Hist(time=list(l,r), event=dementia)\sim certif,
                     data=Paq1000)
```

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idmModel

Generate illness-death model objects

## Description

Function to generate an illness-death model for simulation.

## Usage

```
idmModel(
   scale.illtime = 1/100,
   shape.illtime = 1,
   scale.lifetime = 1/100,
   shape.lifetime = 1,
   scale.waittime = 1/100,
   shape.waittime = 1,
   scale.censtime = 1/100,
   shape.censtime = 1,
   n.inspections = 5,
   schedule = 10,
   punctuality = 5
)
```

#### Arguments

```
scale.illtime
                 Weilbull scale for latent illness time
shape.illtime
                  Weilbull shape for latent illness time
scale.lifetime Weilbull scale for latent life time
shape.lifetime Weilbull shape for latent life time
scale.waittime Weilbull scale for latent life time
shape.waittime Weilbull shape for latent life time
scale.censtime Weilbull scale for censoring time
shape.censtime Weilbull shape for censoring time
                 Number of inspection times
n.inspections
schedule
                 Mean of the waiting time between adjacent inspections.
                  Standard deviation of waiting time between inspections.
punctuality
```

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#### **Details**

Based on the functionality of the lava PACKAGE the function generates a latent variable model (latent illtime, waittime and lifetime) and censoring mechanism (censtime, inspection1,inspection2,...,inspectionK).

The function sim.idmModel then simulates right censored lifetimes and interval censored illness times.

#### Value

A latent variable model object 1vm

#### Author(s)

Thomas Alexander Gerds

```
library(lava)
library(prodlim)
# generate illness-death model based on exponentially
# distributed times
m <- idmModel(scale.illtime=1/70,</pre>
              shape.illtime=1.8,
              scale.lifetime=1/50,
              shape.lifetime=0.7,
              scale.waittime=1/30,
              shape.waittime=0.7)
round(sim(m,6),1)
# Estimate the parameters of the Weibull models
# based on the uncensored exact event times
# and the uncensored illstatus.
set.seed(18)
d <- sim(m,100,latent=FALSE)</pre>
d$uncensored.status <- 1
f <- idm(formula01=Hist(time=illtime,event=illstatus)~1,</pre>
         formula02=Hist(time=lifetime, event=uncensored.status)~1,
         data=d,
         conf.int=FALSE)
print(f)
# Change the rate of the 0->2 and 0->1 transitions
# also the rate of the 1->2 transition
# and also lower the censoring rate
m <- idmModel(scale.lifetime=1/2000,</pre>
              scale.waittime=1/30,
               scale.illtime=1/1000,
               scale.censtime=1/1000)
set.seed(18)
d <- sim(m,50,latent=TRUE)</pre>
d$uncensored.status <- 1
```

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```
f <- idm(formula01=Hist(time=observed.illtime,event=illstatus)~1,</pre>
         formula02=Hist(time=observed.lifetime, event=uncensored.status)~1,
         data=d,
         conf.int=FALSE)
print(f)
# Estimate based on the right censored observations
fc <- idm(formula01=Hist(time=illtime,event=seen.ill)~1,
          formula02=Hist(time=observed.lifetime,event=seen.exit)~1,
          data=d.
          conf.int=FALSE)
print(fc)
# Estimate based on interval censored and right censored observations
fi <- idm(formula01=Hist(time=list(L,R),event=seen.ill)~1,</pre>
          formula02=Hist(time=observed.lifetime,event=seen.exit)~1,
          data=d,
          conf.int=FALSE)
print(fi)
# Estimation of covariate effects:
# X1, X2, X3
m <- idmModel(shape.waittime=2,</pre>
              scale.lifetime=1/2000,
              scale.waittime=1/300,
              scale.illtime=1/10000,
              scale.censtime=1/10000)
distribution(m,"X1") <- binomial.lvm(p=0.3)</pre>
distribution(m,"X2") <- normal.lvm(mean=120,sd=20)</pre>
distribution(m,"X3") <- normal.lvm(mean=50,sd=20)</pre>
regression(m,to="latent.illtime",from="X1") <- 1.7</pre>
regression(m,to="latent.illtime",from="X2") <- 0.07
regression(m,to="latent.illtime",from="X3") <- -0.1
regression(m, to="latent.waittime", from="X1") <- 1.8</pre>
regression(m,to="latent.lifetime",from="X1") <- 0.7</pre>
set.seed(28)
d <- sim(m,100,latent=TRUE)</pre>
table(ill=d$seen.ill,death=d$seen.exit)
# Estimation based on uncensored data
d$uncensored.status <- 1
# uncensored data
F1 <- idm(formula01=Hist(time=illtime,event=illstatus)~X1+X2+X3,
          formula02=Hist(time=lifetime, event=uncensored.status)~X1+X2+X3,
          data=d,conf.int=FALSE)
print(F1)
# Estimation based on right censored data
F2 <- idm(formula01=Hist(time=illtime,event=seen.ill)~X1+X2+X3,
          formula02=Hist(time=observed.lifetime,event=seen.exit)~X1+X2+X3,
          data=d,conf.int=FALSE)
```

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intensity

M-spline estimate of the transition intensity function

## **Description**

M-spline estimate of the transition intensity function and the cumulative transition intensity function for survival and illness-death models

## Usage

```
intensity(times, knots, number.knots, theta, linear.predictor = 0)
```

## **Arguments**

times Time points at which to estimate the intensity function

knots Knots for the M-spline

number.knots Number of knots for the M-splines (and I-splines see details)

theta The coefficients for the linear combination of M-splines (and I-splines see de-

tails)

linear.predictor

Linear predictor beta\*Z. When it is non-zero, transition and cumulative transi-

tion are multiplied by exp(linear.predictor). Default is zero.

#### **Details**

The estimate of the transition intensity function is a linear combination of M-splines and the estimate of the cumulative transition intensity function is a linear combination of I-splines (the integral of a M-spline is called I-spline). The coefficients theta are the same for the M-splines and I-splines.

Important: the theta parameters returned by idm and shr are in fact the square root of the splines coefficients. See examples.

This function is a R-translation of a corresponding Fortran function called susp. susp is used internally by idm and shr.

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#### Value

times The time points at which the following estimates are evaluated.

intensity The transition intensity function evaluated at times.

cumulative.intensity
The cumulative transition intensity function evaluated at times

survival The "survival" function, i.e., exp(-cumulative.intensity)

## Author(s)

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#### See Also

shr, idm

```
data(testdata)
fit.su <- shr(Hist(time=list(l, r), id) ~ cov,</pre>
              data = testdata,method = "Splines",CV = TRUE)
intensity(times = fit.su$time, knots = fit.su$knots,
           number.knots = fit.su$nknots, theta = fit.su$theta^2)
 data(Paq1000)
 fit.idm <- idm(formula02 = Hist(time = t, event = death, entry = e) ~ certif,</pre>
                  formula01 = Hist(time = list(l,r), event = dementia) \sim certif,
                  formula12 = ~ certif, method = "Splines", data = Paq1000)
 # Probability of survival in state 0 at age 80 for a subject with no cep given
 # that he is in state 0 at 70
 su0 <- (intensity(times = 80, knots = fit.idm$knots01,</pre>
                   number.knots = fit.idm$nknots01,
                   theta = fit.idm$theta01^2)$survival
         *intensity(times = 80, knots = fit.idm$knots02,
                   number.knots = fit.idm$nknots02,
                   theta = fit.idm$theta02^2)$survival)/
        (intensity(times = 70, knots = fit.idm$knots01,
                   number.knots = fit.idm$nknots01,
                   theta = fit.idm$theta01^2)$survival
        *intensity(times = 70, knots = fit.idm$knots02,
                   number.knots = fit.idm$nknots02,
                   theta = fit.idm$theta02^2)$survival)
 # Same result as:
 predict(fit.idm, s = 70, t = 80, conf.int = FALSE) # see first element
```

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Paq1000

Paquid data set

## **Description**

Paquid data set composed of 1000 subjects selected randomly from the Paquid data set of 3675 subjects.

#### **Format**

A data frame with 1000 rows and the following 8 columns.

dementia dementia status, 0=non-demented, 1=demented

**death** death status, 0=alive, 1=dead

- e age at entry in the study
- I for demented subjects: age at the visit before the diagnostic visit; for non-demented subjects: age at the last visit (censoring age)
- **r** for demented subjects: age at the diagnostic visit; for non-demented subjects: age at the last visit (censoring age)
- t for dead subjects: age at death; for alive subject: age at the latest news

 $\textbf{certif} \ \ \text{primary school certificate:} \textbf{0} \text{=} \textbf{without certificate}, \textbf{1} \text{=} \textbf{with certificate}$ 

gender gender: 0=female, 1=male

#### **Examples**

data(Paq1000)

plot.idm

Plot method for an illness-death model

## **Description**

Plot estimated baseline transition intensities from an object of class idm optionally with confidence limits.

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## Usage

```
## S3 method for class 'idm'
plot(
 х,
  conf.int = FALSE,
 citype = "shadow",
  add = FALSE,
  axes = TRUE,
  col,
  lwd,
 lty,
 xlim,
 ylim,
 xlab,
 ylab,
 legend = TRUE,
  transition = c("01", "02", "12"),
)
```

# Arguments

x	a idmWeib class object (output from calling idm with the (default) option intensities="Weib".
conf.int	If TRUE show confidence limits
citype	Type of confidence limits, can be "shadow" or "bars"
add	If TRUE add to existing plot
axes	If TRUE axes are drawn
col	Color of the lines
lwd	Width of the lines
lty	Type of the lines
xlim	Limits for x-axis
ylim	Limits for y-axis
xlab	Label for x-axis
ylab	Label for y-axis
legend	If TRUE a legend is drawn, which can be further controlled via SmartControl.
transition	Choose one of the transition intensities: c("01", "02", "12").
	Passed to SmartControl

# Value

Print a plot of the baseline transition intensities of an illness-death model estimated using a Weibull approach.

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## See Also

```
print.idm,summary.idm,idm,
idm
```

## **Examples**

```
library(lava)
library(prodlim)
m <- idmModel(scale.lifetime=1/10,scale.illtime=1/8)
distribution(m,"X") <- binomial.lvm()
regression(m,latent.lifetime~X) <- 0.7
set.seed(30)
d <- sim(m,100)
fit.weib <- idm(formula02=Hist(observed.lifetime,event=seen.exit)~1,
formula01=Hist(time=list(L,R),event=seen.ill)~1,data=d,conf.int=FALSE)
plot(fit.weib)

## FIXME: the limits for the 01 transition are a bit wide!?
## with bootstrap confidence limits
fit.weib <- idm(formula02=Hist(observed.lifetime,event=seen.exit)~1,
formula01=Hist(time=list(L,R),event=seen.ill)~1,data=d,conf.int=TRUE)
plot(fit.weib)</pre>
```

plot.shr

Plot method for a survival model.

## **Description**

Plot estimated baseline survival function from an object of class shr. Pointwise confidence limits are available.

## Usage

```
## S3 method for class 'shr'
plot(
    x,
    type = "shr",
    add = FALSE,
    newdata = NULL,
    cause = NULL,
    col,
    lty,
    lwd,
```

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```
ylim,
xlim,
xlab = "Time",
ylab,
legend = TRUE,
confint = TRUE,
timeOrigin = 0,
axes = TRUE,
percent = TRUE,
...
)
```

# Arguments

X	a shrWeib or a shrSplines class object (output from calling shr function).
type	type of function to plot. The default is "shr".
add	boolean.
newdata	newdata.
cause	cause.
col	col.
lty	lty.
lwd	lwd.
ylim	ylim.
xlim	xlim.
xlab	xlab.
ylab	ylab.
legend	legend.
confint	confint.
timeOrigin	timeOrigin.
axes	axes.
percent	percent.
	other graphical parameters.

# Value

Print a plot of a suvival model.

# Author(s)

R: Celia Touraine < Celia. Touraine@isped.u-bordeaux2.fr> Fortran: Pierre Joly < Pierre. Joly @isped.u-bordeaux2.fr>

## See Also

```
plot.shr
```

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## **Examples**

```
# Weibull survival model
library(prodlim)
data(testdata)
fit.su <- shr(Hist(time=list(l,r),id)~cov,data=testdata)
# pointwise confidence limits
plot(fit.su)
# no pointwise confidence limits
plot(fit.su,confint=FALSE)</pre>
```

predict.idm

Predictions for an illness-death model using either a penalized likelihood approach or a Weibull parametrization.

# Description

Predict transition probabilities and cumulative probabilities from an object of class idmSplines with confidence intervals are calculated.

#### Usage

```
## S3 method for class 'idm'
predict(
  object,
  s,
  t,
  newdata,
  nsim = 200,
  seed = 21,
  conf.int = 0.95,
  lifeExpect = FALSE,
  maxtime,
  ...
)
```

## **Arguments**

object an idm class objects returned by a call to the idm function

s time point at which prediction is made.

t time horizon for prediction.

newdata A data frame with covariate values for prediction.

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nsim	number of simulations for the confidence intervals calculations. The default is 200.
seed	Seed passed to set.seed for Monte Carlo simulation of confidence intervals.
conf.int	Level of confidence, i.e., a value between 0 and 1, the default is 0.95. The default is also used when conf.int=TRUE. To avoid computation of confidence intervals, set conf.int to FALSE or NULL.
lifeExpect	Logical. If TRUE compute life expectancies, i.e., t=Inf.
maxtime	The upper limit of integration for calculations of life expectancies from Weibull parametrizations.
	other parameters.

#### Value

a list containing the following predictions with pointwise confidence intervals:

p00	the transition probability $p_{00}$ .
p01	the transition probability $p_{01}$ .
p11	the transition probability $p_{11}$ .
p12	the transition probability $p_{12}$ .
p02_0	the probability of direct transition from state 0 to state 2.
p02_1	the probability of transition from state 0 to state 2 via state 1.
p02	transition probability $p_{02}$ . Note that p02=p_02_0+p02_1.
F01	the lifetime risk of disease. F01=p01+p02_1.
F0.	the probability of exit from state 0. F0.=p02_0+p01+p02_1.

## Author(s)

R: Celia Touraine < Celia. Touraine@isped.u-bordeaux2.fr> and Thomas Alexander Gerds < tag@biostat.ku.dk> Fortran: Pierre Joly < Pierre. Joly @isped.u-bordeaux2.fr>

#### See Also

idm

18 print.idm

print.idm

Print method for idm objects

#### **Description**

Print a summary of a fitted illness-death model

## Usage

```
## S3 method for class 'idm'
print(x, conf.int = 0.95, digits = 4, pvalDigits = 4, eps = 0.0001, ...)
```

#### Arguments

x Class idm object, i.e. the result of a call to the idm function with intensities="Weib".

conf.int The level of confidence for the hazard ratios. The default is 0.95.

digits Number of digits to print.

pvalDigits Number of digits to print for p-values.

eps Passed to format.pval.

... Not used.

#### Value

No return value.

print.shr 19

#### Author(s)

Celia Touraine <Celia.Touraine@isped.u-bordeaux2.fr>, Thomas A. Gerds <tag@biostat.ku.dk>

#### See Also

```
summary.idm, plot.idm
```

## **Examples**

print.shr

Print method for shrSplines objects

# Description

Print a summary of a fitted illness-death model using the penalized likelihood approach.

## Usage

```
## S3 method for class 'shr'
print(x, conf.int = 0.95, digits = 4, pvalDigits = 4, eps = 0.0001, ...)
```

## **Arguments**

```
x a shr object, i.e., the result of a call to the shr function with hazard="Splines".

conf.int The level of confidence for the hazard ratios. The default is 0.95.

digits number of digits to print.

pvalDigits number of digits to print for p-values.

eps convergence criterion used for p-values.

other unusued arguments.
```

#### Value

No return value.

20 shr

#### Author(s)

R: Celia Touraine < Celia. Touraine@isped.u-bordeaux2.fr> Fortran: Pierre Joly < Pierre. Joly@isped.u-bordeaux2.fr>

#### See Also

```
summary.shr, plot.shr
```

## **Examples**

```
# a penalized survival model
library(prodlim)
data(testdata)
fit.su <- shr(Hist(time=list(1,r),id)~cov,data=testdata,method="Splines")
print(fit.su)</pre>
```

shr

Fit a survival model

## **Description**

Fit a survival model using either a semi-parametric approach (penalized likelihood with an approximation of the hazard function by linear combination of M-splines) or a parametric approach (specifying a Weibull distribution on the hazard function). Left-truncated, right-censored, and intervalcensored data are allowed.

## Usage

```
shr(
  formula,
  data,
  eps = c(5, 5, 3),
  n.knots = 7,
  knots = "equidistant",
  CV = FALSE,
  kappa = 10000,
  conf.int = 0.95,
  maxiter = 200,
  method = "Weib",
  print.iter = FALSE,
  na.action = na.omit
)
```

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#### **Arguments**

formula a formula object with the response on the left hand side and the terms on the right hand side. The response must be a survival object or Hist object as returned by the 'Surv' or 'Hist' function. data a data frame in which to interpret the variables named in the formula. a vector of length 3 for the convergence criteria (criterion for parameters, crieps terion for likelihood, criterion for second derivatives). The default is c(5,5,3)and corresponds to criteria equals to  $10^{-5}$ ,  $10^{-5}$  and  $10^{-3}$ . n.knots Argument only active for the penalized likelihood approach method="splines". Number of knots for the splines to use to approximate the hazard function. The default is 7. If knots are given as a vector this argument is ignored. The algorithm needs least 5 knots and at most 20 knots. Argument only active for the penalized likelihood approach method="splines". knots There are three ways to control the placement of the knots between the smallest and the largest of all time points: knots="equidistant" Knots are placed with same distance on the time scale. knots="quantiles" Knots are placed such that the number of observations is roughly the same between knots. **knots=list()** List of length 3. The list elements are the actual placements (timepoints) of the knots for the M-spline. The algorithm reugires at least 5 knots and allows no more than 20 knots. C۷ binary variable equals to 1 when search (by approximated cross validation) of the smoothing parameter kappa and 0 otherwise. Argument for the penalized likelihood approach. The default is 0. Argument only active for the penalized likelihood approach method="splines". kappa A positive number (smoothing parameter) If CV=1 the value is used as a starting value for a cross validation search to optimize kappa. conf.int Level of confidence pointwise confidence intervals of the survival and hazard functions, i.e., a value between 0 and 1, the default is 0.95. The default is also used when conf.int=TRUE. To avoid computation of confidence intervals, set conf.int to FALSE or NULL. maxiter maximum number of iterations. The default is 200. method type of estimation method: "Splines" for a penalized likelihood approach with approximation of the hazard function by M-splines, "Weib" for a parametric approach with a Weibull distribution on the hazard function. Default is "Weib". print.iter boolean parameter. Equals to TRUE to print the likelihood during the iteration process, FALSE otherwise. Default is FALSE. This option is not running on Winhow NAs are treated. The default is first, any na.action attribute of data, second na.action a na.action setting of options, and third 'na.fail' if that is unset. The 'factoryfresh' default is na.omit. Another possible value is NULL.

## Details

The estimated parameters are obtained using the robust Marquardt algorithm (Marquardt, 1963) which is a combination between a Newton-Raphson algorithm and a steepest descent algorithm.

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#### Value

call

**coef** regression parameters.

loglik vector containing the log-likelihood without and with covariate.

modelPar Weibull parameters.

N number of subjects.

NC number of covariates.

nevents number of events.

modelResponse model response: Hist or Surv object.

converged integer equal to 1 when the model converged, 2, 3 or 4 otherwise.

time times for which survival and hazard functions have been evaluated for plotting.

hazard matched values of the hazard function.

lowerHazard lower confidence limits for hazard function.

upperHazard upper confidence limits for hazard function.

surv matched values of the survival function.

lowerSurv lower confidence limits for survival function.

upperSurv upper confidence limits for survival function.

**RR** vector of relative risks.

V variance-covariance matrix.

se standard errors.

knots knots of the M-splines estimate of the hazard function.

**nknots** number of knots.

**CV** a binary variable equals to 1 when search of the smoothing parameter kappa by approximated cross-validation, 1 otherwise. The default is 0.

**niter** number of iterations.

cv vector containing the convergence criteria.

**na.action** observations deleted if missing values.

#### Author(s)

R: Celia Touraine <celia.touraine@icm.unicancer.fr>Fortran: Pierre Joly <Pierre.Joly@isped.u-bordeaux2.fr>

## References

D. Marquardt (1963). An algorithm for least-squares estimation of nonlinear parameters. *SIAM Journal of Applied Mathematics*, 431-441.

#### See Also

```
shr, print.shr, summary.shr, print.shr,
```

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#### **Examples**

sim.idmModel

Simulate illness-death model data

## **Description**

Function to simulate illness-death model data

## Usage

```
## $3 method for class 'idmModel'
sim(
    x,
    n,
    illness.known.at.death = TRUE,
    compliance = 1,
    latent = FALSE,
    keep.inspectiontimes = FALSE,
    ...
)
```

#### Arguments

24 sim.survIC

```
keep.inspectiontimes
if TRUE keep the inspection times.
... Extra arguments given to sim
```

#### **Details**

Based on the functionality of the lava PACKAGE

## Value

A data set with interval censored observations from an illness-death model

### Author(s)

Thomas Alexander Gerds

# **Examples**

```
example(idmModel)
help(idmModel)
```

sim.survIC

Simulate interval censored survival data

## **Description**

Function to simulate interval censored survival data

# Usage

```
## S3 method for class 'survIC'
sim(x, n, compliance = 1, latent = TRUE, keep.inspectiontimes = FALSE, ...)
```

#### **Arguments**

x An survIC object as obtained with survIC

n Number of observations

compliance Probability of missing an inspection time.

latent if TRUE keep the latent event times

keep.inspectiontimes

if TRUE keep the inspection times.

... Extra arguments given to sim

## **Details**

Based on the functionality of the lava PACKAGE

simulateIDM 25

## Value

A data set with interval censored observations

#### Author(s)

Thomas Alexander Gerds

## **Examples**

```
library(lava)
example(survIC)
help(survIC)
ol <- survIC()
dat.ol <- sim(ol,10)</pre>
```

simulateIDM

Sample illness-death model data

## **Description**

Simulate data from an illness-death model with interval censored event times and covariates

#### Usage

```
simulateIDM(n = 100)
```

#### **Arguments**

n

number of observations

#### **Details**

Simulate data from an illness-death model with interval censored event times and covariates for the purpose of illustrating the help pages of the SmoothHazard package. See the body of the function for details, i.e., evaluate simulateIDM

## Value

Object with class data. frame which contains the simulated data.

## See Also

idmModel sim.idmModel

```
# simulateIDM
simulateIDM(100)
```

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summary.idm

Summary of a fitted illness-death model

#### **Description**

Summarize the event history data of an illness-death regression model and show regression coefficients for transition intensities

#### Usage

```
## S3 method for class 'idm'
summary(object, conf.int = 0.95, digits = 4, pvalDigits = 4, eps = 0.0001, ...)
```

## **Arguments**

object a idmSplines object, i.e., the result of a call to the idm function with intensities="Splines".

conf.int The level of confidence for the hazard ratios. The default is 0.95.

digits number of digits to print.

pvalDigits number of digits to print for p-values.

eps convergence criterion used for p-values.

... other unusued arguments.

#### Value

No return value.

## Author(s)

R: Celia Touraine < Celia. Touraine @isped.u-bordeaux2.fr> Fortran: Pierre Joly < Pierre. Joly @isped.u-bordeaux2.fr>

#### See Also

```
idm, print.idm, plot.idm
```

summary.shr 27

summary.shr	Summary of a fitted survival model using a penalized likelihood approach

## **Description**

Print a short summary of a fitted illness-death model using the penalized likelihood approach.

## Usage

```
## S3 method for class 'shr'
summary(object, conf.int = 0.95, digits = 4, pvalDigits = 4, eps = 0.0001, ...)
```

## **Arguments**

object a shr object, i.e., the result of a call to the shr function.

conf.int The level of confidence for the hazard ratios. The default is 0.95.

digits number of digits to print.

pvalDigits number of digits to print for p-values.

eps convergence criterion used for p-values.

other unusued arguments.

#### Value

No return value.

## Author(s)

Celia Touraine < Celia. Touraine@isped.u-bordeaux2.fr>

## See Also

```
shr, print.shr, plot.shr
```

```
# a penalized survival model
data(testdata)
library(prodlim)
fit.su <- shr(Hist(time=list(1,r),id)~cov,data=testdata,method="Splines")
summary(fit.su)

# Weibull survival model
data(testdata)
fit.su <- shr(Hist(time=list(1,r),id)~cov,data=testdata)
summary(fit.su)</pre>
```

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survIC

Generate survival model objects

### Description

Function to generate a latent variable model for interval censored survival times.

#### Usage

```
survIC(
   scale.time = 1/100,
   shape.time = 1,
   n.inspections = 5,
   schedule = 10,
   punctuality = 5
)
```

## **Arguments**

scale.time Weilbull scale for latent time shape.time Weilbull shape for latent time n.inspections Number of inspection times

schedule Mean of the waiting time between adjacent inspections.

punctuality Standard deviation of waiting time between inspections.

#### **Details**

Based on the functionality of the lava PACKAGE the function generates a latent variable model with a latent time and a censoring mechanism (censtime, inspection1,inspection2,...,inspectionK).

The function sim. survIC then simulates interval censored times.

#### Value

A latent variable model object 1vm

#### Author(s)

Thomas Alexander Gerds

```
library(lava)
library(prodlim)
# generate survival model based on exponentially
# distributed times
m <- survIC(scale.time=1/50, shape.time=0.7)</pre>
```

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testdata

Data set for survival models: right-censored and interval-censored data.

# Description

A simulated data frame for survival models composed of right-censored and interval-censored data.

#### **Format**

A data frame with 936 observations on the following 4 variables.

- 1 for diseased subjects: left endpoint of censoring interval; for non-diseased subjects: right censoring time
- ${f r}$  for diseased subjects: right endpoint of censoring interval; for non-diseased subjects: right censoring time for the disease event

id disease status

cov covariate

```
data(testdata)
head(testdata)
```

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