# Package ‘survsim’

October 16, 2015

- **Type**: Package
- **Title**: Simulation of Simple and Complex Survival Data
- **Version**: 1.1.4
- **Date**: 2015-10-07
- **Encoding**: UTF-8
- **Author**: David Moriña (Centre for Research in Environmental Epidemiology, CREAL) and Albert Navarro (Universitat Autònoma de Barcelona)
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- **Description**: Simulation of simple and complex survival data including recurrent and multiple events and competing risks.
- **Depends**: R (>= 2.13.1), eha, statmod
- **License**: GPL (>= 2)
- **NeedsCompilation**: no
- **Repository**: CRAN
- **Date/Publication**: 2015-10-16 17:29:26

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Description

Simulation of cohorts in a context of simple and complex survival analysis, multiple events and recurrent events including several covariates, individual heterogeneity and periods at risk before and after the initial time of follow-up.

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<td>Log-normal</td>
<td>( 1 - \Theta((\log(t) - \mu)/\sigma) )</td>
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<td>Log-logistic</td>
<td>( 1/(1 + (\lambda t)^{1/\gamma}) )</td>
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Where \( \Theta \) is the standard normal cumulative distribution.

In order to simulate censored survival data, two survival distributions are required, one for the uncensored survival times that would be observed if the follow-up had been sufficiently long to reach the event and another representing the censoring mechanism. The uncensored survival distribution, \( T'_i \), for \( i = 1, \ldots, n \) subjects, could be generated to depend on a set of covariates with a specified relationship with survival, which represents the true prognostic importance of each covariate (Burton, 2006). The package allows to simulate times by means of using Weibull (and exponential as a particular case), log-normal and log-logistic distributions, as such is showed in previous table. To induce individual heterogeneity or within-subject correlation we generate \( Z_i \), a random effect covariate that follows a particular distribution (Uniform or Normal).

\[
t_i = t'_i \cdot z_i
\]

Where \( z_i = 1 \), for all subjects, we are in the case of individual homogeneity and the survival times are completely specified by the covariates. Random non-informative right censoring, \( C_i \), can be generated in a similar manner to the uncensored survival times, \( T'_i \), by assuming a particular distribution for the censoring times (previous table), but without including any covariates nor individual heterogeneity. The observation times, \( Y'_i \), incorporating both events and censored observations are calculated for each case by combining the uncensored survival times, \( T_i \), and the censoring times, \( C_i \). If the uncensored survival time for an observation is less than or equal to the censored time, then the event is considered to be observed and the observation time equals the uncensored survival time, otherwise the event is considered censored and the observation time equals the censored time. In other words, once simulated \( t_i \) and \( c_i \), we can define \( Y'_i = \min(t_i, c_i) \) as the obervation time.
with $\delta_i$ an indicator of non-censoring, i.e. $\delta_i = I(t_i \leq c_i)$. While all $y_i'$ start at 0, the package allows create dynamic cohorts. We can generate entry times higher than 0 adding a $t_0$ value corresponding with an uniform distribution in $[0,t_{\text{max follow-up}}]$. We can also simulate subjects at risk before of the initial time of follow-up ($y_i' = 0$), by including an uniform distribution for $t_0$ between $[-t_{\text{max old}},0)$ for a fixed percentage of subjects. Then:

$$y_i = y_i' + t_0$$

where $t_0$ follows a uniform distribution in $[0,t_{\text{max follow-up}}]$ if entry time is 0 or more and $t_0$ is uniform distributed in $[-t_{\text{max old}},0)$ if entry time is less than 0. Therefore, $t_0$ represents the initial point of the episode, $y_i$ the endpoint and $y_i'$ is the length. Note that $y_i' + t_0$ can be higher than $t_{\text{max follow-up}}$, and in this case $y_i$ will be set at $t_{\text{max follow-up}}$ and $\delta_i = 0$. The observations corresponding to the subjects at risk before of the initial time of follow-up have $t_0$ negative, then the initial point of the episode will be set at 0. $y_i$ may also be negative, in this case this episode will not be included in the simulated data, as long as this episode won’t be observed in practice.

Details

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The package provide a tool for simulation of cohorts in a simple single-event context through the function `simpleNsurvNsim`, in a recurrent event context with the function `rec.ev.sim`, in a multiple event context with the function `mult.ev.sim` and in a competing risks context with the function `crisk.sim`, and it also allows the user to generate aggregated data from the simulated cohort, by means of the function `accum`.

Author(s)

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References


Reis RJ, Utzet M, La Rocca PF, Nedel FB, Martin M, Navarro A. Previous sick leaves as predictor of subsequent ones. Int Arch Occup Environ Health 2011 Jun;84(5):491-499.


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**accum**

*Aggregate data from a simulated cohort.*

**Description**

Aggregate the observed number of events suffered by a subject, the time of follow-up, the duration of all the observed episodes and the real number of events suffered in all subject history.

**Usage**

```r
accum(data)
```

**Arguments**

- `data` An object of class `mult.ev.data.sim`, if the individual cohort has been simulated in a multiple event situation or an object of class `rec.ev.data.sim`, if the individual cohort has been simulated in a recurrent event situation. Note that, although the routine will work, it’s probably not much useful in other contexts than recurrent event situation.

**Details**

The output contains `z` and `real.ep.accum` because they can be interesting when analyzing several aspects as missing data or individual heterogeneity, although those variables cannot be observed in a real cohort.

**Value**

An object of class `sim.ev.agg.data`. It is a data frame with a row for each subject in `data`, and the following columns

- `nid` an integer number that identifies the subject.
- `old` real value indicating the time that the individual was at risk before the beginning of the follow-up.
`accum`

- `risk.bef`: Boolean indicating if the subject was at risk before the beginning of the follow-up time or not.
- `z`: individual heterogeneity, generated according to the specified distribution.
- `x`: value of each covariate randomly generated for each subject in the cohort.
- `obs.ep.accum`: aggregated number of episodes suffered by an individual since the beginning of subject’s follow-up time.
- `real.ep.accum`: aggregated number of episodes suffered by an individual from the beginning of subject’s history.
- `time.accum`: global time of follow-up for each individual.
- `long.accum`: global time not at risk within the follow-up time, corresponding to the sum of times between the end of an event and the beginning of the next.

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**References**


Reis RJ, Utzet M, La Rocca PF, Nedel FB, Martin M, Navarro A. Previous sick leaves as predictor of subsequent ones. Int Arch Occup Environ Health 2011 Jun;84(5):491-499.


**See Also**

`rec.ev.sim`, `mult.ev.sim`, `crisk.sim`, `survsim`, `simple.surv.sim`

**Examples**

```r
### A cohort with 500 subjects, with a maximum follow-up time of 1825 days and
### just a covariate, following a Bernoulli distribution, and a corresponding
### beta of -0.4, -0.5, -0.6 and -0.7 for each episode, in a context of recurrent
### events.

sim.data <- rec.ev.sim(n=500, foltime=1825, dist.ev=c('lnorm', 'llogistic', 'weibull', 'weibull'), anc.ev=c(1.498, 0.924, 0.923, 1.051), beta.ev=c(7.195, 6.583, 6.678, 6.430), anc.cens=c(1.272, 1.218, 1.341, 1.484), beta.cens=c(7.315, 6.975, 6.712, 6.399),
```
### crisk.sim

#### Generate a cohort in a competing risks context

**Description**

Simulation of cohorts in a context of competing risks survival analysis including several covariates, individual heterogeneity and periods at risk prior and after the start of follow-up.

Competing risks analysis considers time-to-first-event and the event type, possibly subject to right censoring (Beyersmann et al., 2009)

**Usage**

```r
crisk.sim(n, foltime, dist.ev, anc.ev, beta0.ev, dist.cens="weibull", anc.cens, beta0.cens, z=NULL, beta=NA, x=NA, nsit)
```

**Arguments**

- `n`: integer value indicating the desired size of the cohort to be simulated.
- `foltime`: real number that indicates the maximum time of follow-up of the simulated cohort.
- `dist.ev`: vector of arbitrary size indicating the time to event distributions, with possible values `weibull` for the Weibull distribution, `lnorm` for the log-normal distribution and `llogistic` for the log-logistic distribution.
- `anc.ev`: vector of arbitrary size of real components containing the ancillary parameters for the time to event distributions.
- `beta0.ev`: vector of arbitrary size of real components containing the $\beta_0$ parameters for the time to event distributions.
- `dist.cens`: string indicating the time to censoring distributions, with possible values `weibull` for the Weibull distribution, `lnorm` for the log-normal distribution, `llogistic` for the log-logistic distribution and `unif` for the uniform distribution. If no distribution is introduced, the time to censoring is assumed to follow a Weibull distribution.
- `anc.cens`: real number containing the ancillary parameter for the time to censoring distribution or the maximum in case of uniform distributed time to censoring.
- `beta0.cens`: real number containing the $\beta_0$ parameter for the time to censoring distribution or the minimum in case of uniform distributed time to censoring.
list of vectors with three elements containing information relative to a random
effect used in order to introduce individual heterogeneity. Each vector in the
list refers to a possible competing risk, so the number of vectors must be equal
to nsit or equal to 1 if the same random effect will be used for all the causes.
The first element indicates the distribution: "unif" states for a uniform distri-
bution, "gamma" states for a gamma distribution, "exp" states for an exponential
distribution, "weibull" states for a Weibull distribution and "invgauss" states
for an inverse gaussian distribution. The second and third elements indicates
the minimum and maximum in the case of a uniform distribution (both must be
positive) and the parameters in the case of the rest of distributions. Note that
just one parameter is needed in the case of exponential distribution. Its default
value is NULL, indicating that no individual heterogeneity is introduced.

list of vectors indicating the effect of the corresponding covariate. The number
of vectors in beta must match the number of covariates, and the length of each
vector must match the number of events considered. Its default value is NA,
indicating that no covariates are included.

list of vectors indicating the distribution and parameters of any covariate that the
user needs to introduce in the simulated cohort. The possible distributions are
"normal" for normal distribution, "unif" for uniform distribution and "bern" for Bernoulli distribution. Its default value is NA, indicating that no covariates
are included. The number of vectors in x must match the number of vectors
in beta. Each vector in x must contain the name of the distribution and the
parameter(s), which are: the probability of success in the case of a Bernoulli
distribution; the mean and the variance in the case of a normal distribution; and
the minimum and maximum in the case of a uniform distribution.

Number of different events that a subject can suffer. It must match the number
distributions specified in dist.ev.

In order to get the function to work properly, the length of the vectors containing the parameters of
the time to event and the number of distributions indicated in the parameter dist.ev must be the
same.

An object of class mult.ev.data.sim. It is a data frame containing the events suffered by the
corresponding subjects. The columns of this data frame are detailed below

an integer number that identifies the subject.
cause cause of the event corresponding to the follow-up time of the individual.
time time until the corresponding event happens (or time to subject drop-out).
status logical value indicating if the corresponding event has been suffered or not.
start time at which the follow-up time begins for each event.
stop time at which the follow-up time ends for each event.
z Individual heterogeneity generated according to the specified distribution.
x value of each covariate randomly generated for each subject in the cohort.
Author(s)

David Moriña, Centre for Research in Environmental Epidemiology (CREAL) and Albert Navarro, Universitat Autònoma de Barcelona

References


See Also

survsim-package, accum, rec.ev.sim, mult.ev.sim, simple.surv.sim

Examples

### A cohort with 50 subjects, with a maximum follow-up time of 100 days and two covariates, following Bernoulli distributions, and a corresponding beta of 0.1698695 and 0.0007010932 for each event for the first covariate and a corresponding beta of 0.3735146 and 0.5591244 for each event for the second covariate. Notice that the time to censorship is assumed to follow a log-normal distribution.

```r
sim.data <- crisk.sim(n=50, foltime=100, dist.ev=c("lnorm","lnorm"), anc.ev=c(1.497687, 0.5268302),beta.ev=c(3.80342, 2.535374),dist.cens="lnorm", anc.cens=1.242733,beta0.cens=5.421748,z=list(c("unif", 0.8,1.2), c("unif", 0.9, 1.5)), beta=list(c(0.1698695,0.0007010932),c(0.3735146,0.5591244)), x=list(c("bern", 0.381), c("bern", 0.564)), nsit=2)
summary(sim.data)
```

---

**Generate a cohort with multiple events**

### Description

Simulation of cohorts in a context of multiple event survival analysis including several covariates, individual heterogeneity and periods at risk prior and after the start of follow-up.

Multiple event data occurs when each subject can have more than one event of entirely different natures (Kelly, 2000). Examples of this type of event are the occurrence of tumours at different sites in the body or multiple sequelae after a surgery.

We can obtain the observation time of the $k$-th event in the $i$-th subject, $y_{ik}$, in the same manner that we can simulate $k$ simple independent survival data. Notice that, in multiple-type events, $T_{ik}$ and $C_i$ are mutually independent and, furthermore, the failure in each event is independent of the others (within each subject all $y_{ik}$ are independents for all $k$).
**Usage**

```r
dist.ev = c("weibull", "log-normal", "log-logistic")
anc.ev = c(0.1, 0.2)
beta0.ev = c(0.5, 0.6)
```

**Arguments**

- `n`: integer value indicating the desired size of the cohort to be simulated.
- `foltime`: real number that indicates the maximum time of follow-up of the simulated cohort.
- `dist.ev`: vector of arbitrary size indicating the time to event distributions, with possible values `weibull` for the Weibull distribution, `lnorm` for the log-normal distribution and `llogistic` for the log-logistic distribution.
- `anc.ev`: vector of arbitrary size of real components containing the ancillary parameters for the time to event distributions.
- `beta0.ev`: vector of arbitrary size of real components containing the $\beta_0$ parameters for the time to event distributions.
- `dist.cens`: string indicating the time to censoring distributions, with possible values `weibull` for the Weibull distribution, `lnorm` for the log-normal distribution, `llogistic` for the log-logistic distribution and `unif` for the uniform distribution. If no distribution is introduced, the time to censoring is assumed to follow a Weibull distribution.
- `anc.cens`: real number containing the ancillary parameter for the time to censoring distribution or the maximum in case of uniform distributed time to censoring.
- `beta0.cens`: real number containing the $\beta_0$ parameter for the time to censoring distribution or the minimum in case of uniform distributed time to censoring.
- `z`: list of vectors with three elements containing information relative to a random effect used in order to introduce individual heterogeneity. Each vector in the list refers to a possible event, so the number of vectors must be equal to `nsit` or equal to 1 if the same random effect will be used for all the events. The first element indicates the distribution: "unif" states for a uniform distribution, "gamma" states for a gamma distribution, "exp" states for an exponential distribution, "weibull" states for a Weibull distribution and "invgauss" states for an inverse gaussian distribution. The second and third elements indicates the minimum and maximum in the case of a uniform distribution (both must be positive) and the parameters in the case of the rest of distributions. Note that just one parameter is needed in the case of exponential distribution. Its default value is NULL, indicating that no individual heterogeneity is introduced.
- `beta`: list of vectors indicating the effect of the corresponding covariate. The number of vectors in `beta` must match the number of covariates, and the length of each vector must match the number of events considered. Its default value is NA, indicating that no covariates are included.
- `x`: list of vectors indicating the distribution and parameters of any covariate that the user needs to introduce in the simulated cohort. The possible distributions are "normal" for normal distribution, "unif" for uniform distribution and "bern" for Bernoulli distribution. Its default value is NA, indicating that no covariates are included.
are included. The number of vectors in \( x \) must match the number of vectors in \( \beta \). Each vector in \( x \) must contain the name of the distribution and the parameter(s), which are: the probability of success in the case of a Bernoulli distribution; the mean and the variance in the case of a normal distribution; and the minimum and maximum in the case of a uniform distribution.

\texttt{nsit}  
Number of different events that a subject can suffer. It must match the number of distributions specified in \texttt{dist.ev}.

**Details**

In order to get the function to work properly, the length of the vectors containing the parameters of the time to event and the number of distributions indicated in the parameter \texttt{dist.ev} must be the same.

**Value**

An object of class \texttt{mult.ev.data.sim}. It is a data frame containing the events suffered by the corresponding subjects. The columns of this data frame are detailed below:

\begin{itemize}
  \item \texttt{nid}  
an integer number that identifies the subject.
  \item \texttt{ev.num}  
  number of the event corresponding to the follow-up time of the individual.
  \item \texttt{time}  
  time until the corresponding event happens (or time to subject drop-out).
  \item \texttt{status}  
  logical value indicating if the corresponding event has been suffered or not.
  \item \texttt{start}  
  time at which the follow-up time begins for each event.
  \item \texttt{stop}  
  time at which the follow-up time ends for each event.
  \item \texttt{z}  
  Individual heterogeneity generated according to the specified distribution.
  \item \texttt{x}  
  value of each covariate randomly generated for each subject in the cohort.
\end{itemize}

**Author(s)**

David Moriña, Centre for Research in Environmental Epidemiology (CREAL) and Albert Navarro, Universitat Autònoma de Barcelona

**References**


**See Also**

\texttt{survsim-package, accum, rec.ev.sim, crisk.sim, simple.surv.sim}
Examples

A cohort with 1000 subjects, with a maximum follow-up time of 3600 days and two covariates, following a Bernoulli and uniform distribution respectively, and a corresponding beta of -0.4, -0.5, -0.6 and -0.7 for each event for the first covariate and a corresponding beta of 0, 0, 0 and 1 for each event for the second covariate. Notice that the time to censorship is assumed to follow a Weibull distribution, as no other distribution is stated and random effect is the same for all events.

```
sim.data <- mult.ev.sim(n=1000, foltime=3600, dist.ev=c('llogistic', 'weibull', 'weibull', 'weibull'), anc.ev=c(0.69978200185280, 0.7969165913027, 0.82218416457321, 0.85817155198598), beta0.ev=c(5.84298525742252, 5.94362650003287, 5.7018252804637, 5.46865223339755), anc.cens=1.1770368709519, beta0.cens=7.39773677281100, z=list(c("unif", 0.8,1.2)), beta=list(c(-0.4,-0.5,-0.6,-0.7), c(0,0,0,1)), x=list(c("bern", 0.5), c("unif", 0.7, 1.3)), nsit=4)
```

```
sim.data <- mult.ev.sim(n=1000, foltime=3600, dist.ev=c('llogistic', 'weibull', 'weibull', 'weibull'), anc.ev=c(0.69978200185280, 0.7969165913027, 0.82218416457321, 0.85817155198598), beta0.ev=c(5.84298525742252, 5.94362650003287, 5.7018252804637, 5.46865223339755), anc.cens=1.1770368709519, beta0.cens=7.39773677281100, z=list(c("unif", 0.8,1.2)), beta=list(c(-0.4,-0.5,-0.6,-0.7), c(0,0,0,1)), x=list(c("bern", 0.5), c("unif", 0.7, 1.3)), nsit=4)
```

```
sim.data <- mult.ev.sim(n=1000, foltime=3600, dist.ev=c('llogistic', 'weibull', 'weibull', 'weibull'), anc.ev=c(0.69978200185280, 0.7969165913027, 0.82218416457321, 0.85817155198598), beta0.ev=c(5.84298525742252, 5.94362650003287, 5.7018252804637, 5.46865223339755), anc.cens=1.1770368709519, beta0.cens=7.39773677281100, z=list(c("unif", 0.8,1.2)), beta=list(c(-0.4,-0.5,-0.6,-0.7), c(0,0,0,1)), x=list(c("bern", 0.5), c("unif", 0.7, 1.3)), nsit=4)
```

summary(sim.data)
```

### Generate a cohort with recurrent events

**Description**

Simulation of cohorts in a context of recurrent event survival analysis including several covariates, individual heterogeneity and periods at risk before and after the initial time of follow-up.

Recurrent event data is a type of multiple event where the subject can experience repeated occurrences of the same type (Kelly, 2000), for example repeated asthma attacks or sick leave episodes. In practice, the hazard of an recurrent event can vary depending on the number of previous occurrences, in terms of shape and intensity (Reis, 2011; Navarro, 2012). However, simulations based on a mixture of distributions with different baseline hazard rates are quite rare (Bender, 2005; Metcalfe, 2006).

In a recurrent data context, each subject can present different number of episodes. We talk of episodes (or occurrences) rather than events since each occurrence is a new episode of the same event. This package supposes that there exists one different and independent \( Y_k \) distribution depending on \( k \), the number of episode at risk. The simulating process for each \( Y_k \) is the same than for the multiple events situation (see `mult.ev.sim`), but in this case, obviously, a subject cannot be at risk for the \( k \)-th episode if he/she hadn’t had the \( k - 1 \)-th.

**Usage**

```
rec.ev.sim(n, foltime, dist.ev, anc.ev, beta0.ev, dist.cens=rep("weibull", length(beta0.cens)), anc.cens, beta0.cens, z=NULL, beta=NA, x=NA, lambda=NA, max.ep=Inf, priskb=0, max.old=0)
```
Arguments

- **n**: integer value indicating the desired size of the cohort to be simulated.
- **foltime**: real number that indicates the maximum time of follow-up of the simulated cohort.
- **dist.ev**: vector of arbitrary size indicating the time to event distributions, with possible values weibull for the Weibull distribution, lnorm for the log-normal distribution and llogistic for the log-logistic distribution. If a subject suffers more episodes than specified distributions, the last distribution specified here is used to generate times corresponding to posterior episodes.
- **anc.ev**: vector of arbitrary size of real components containing the ancillary parameters for the time to event distributions.
- **beta0.ev**: vector of arbitrary size of real components containing the $\beta_0$ parameters for the time to event distributions.
- **dist.cens**: string indicating the time to censoring distributions, with possible values weibull for the Weibull distribution, lnorm for the log-normal distribution, llogistic for the log-logistic distribution and unif for the uniform distribution. If no distribution is introduced, the time to censoring is assumed to follow a Weibull distribution.
- **anc.cens**: real number containing the ancillary parameter for the time to censoring distribution or the maximum in case of uniform distributed time to censoring.
- **beta0.cens**: real number containing the $\beta_0$ parameter for the time to censoring distribution or the minimum in case of uniform distributed time to censoring.
- **z**: list of vectors with three elements containing information relative to a random effect used in order to introduce individual heterogeneity. Each vector in the list refers to a possible episode, so the number of vectors must be equal to nsit or equal to 1 if the same random effect will be used for all the episodes. The first element indicates the distribution: "unif" states for a uniform distribution, "gamma" states for a gamma distribution, "exp" states for an exponential distribution, "weibull" states for a Weibull distribution and "invgauss" states for an inverse gaussian distribution. The second and third elements indicate the minimum and maximum in the case of a uniform distribution (both must be positive) and the parameters in the case of the rest of distributions. Note that just one parameter is needed in the case of exponential distribution. Its default value is NULL, indicating that no individual heterogeneity is introduced.
- **beta**: list of vectors indicating the effect of the corresponding covariate. The number of vectors in beta must match the number of covariates, and the length of each vector must match the number of events considered. Its default value is NA, indicating that no covariates are included.
- **x**: list of vectors indicating the distribution and parameters of any covariate that the user need to introduce in the simulated cohort. The possible distributions are "normal" for a normal distribution, "unif" for a uniform distribution and "bern" for a Bernoulli distribution. Its default value is NA, indicating that no covariates are included. The number of vectors in x must match the number of vectors in beta. Each vector in x must contain the name of the distribution and the parameter(s), which are: the probability of success in the case of a Bernoulli
distribution; the mean and the variance in the case of a normal distribution; and the minimum and maximum in the case of a uniform distribution.

lambda
real number indicating the mean duration of each event or discontinuous risk time, assumed to follow a zero-truncated Poisson distribution. Its default value is NA, corresponding to the case where the duration of each event or discontinuous risk time is unnecessary information for the user.

max.ep
integer value that matches the maximum permitted number of episodes per subject. Its default value is Inf, i.e. the number of episodes per subject is no limited.

priskb
proportion of subjects at risk prior to the start of follow-up, defaults to 0.

max.old
maximum time at risk prior to the start of follow-up.

Details
In order to get the function to work properly, the length of the vectors containing the parameters of the time to event and time to censure distributions and the number of distributions indicated in the parameter dist must be the same. Finally, priskb and max.old must be positive numbers, with priskb being between 0 and 1. Notice that large values of max.old can result in the routine taking a long time to simulate a cohort with the specified size.

Value
An object of class rec.ev.data.sim. It is a data frame containing the episodes suffered by the corresponding subjects. The columns of the data frame are detailed below

nid
an integer number that identifies the subject.
real.episode
number of the episode corresponding to the real history of the individual.
obs.episode
number of the episode corresponding to the follow-up time of the individual.
time
time until the corresponding event happens (or time to subject drop-out), regarding the beginning of the follow-up time.
status
logical value indicating if the episode corresponds to an event or a drop-out.
start
time at which an episode starts, taking the beginning of follow-up as the origin of the time scale.
stop
time at which an episode ends, taking the beginning of follow-up as the origin of the time scale.
time2
time until the corresponding event happens (or time to subject drop-out), in calendar time.
start2
time at which an episode starts, where the time scale is calendar time.
stop2
time at which an episode ends, where the time scale is calendar time.
old
real value indicating the time that the individual was at risk before the beginning of follow-up.
risk.bef
factor that indicates if an individual was at risk before the beginning of follow-up or not.
long
time not at risk immediately after an episode.
z
Individual heterogeneity generated according to the specified distribution.
x
value of each covariate randomly generated for each subject in the cohort.
Author(s)

David Moriña, Centre for Research in Environmental Epidemiology (CREAL) and Albert Navarro, Universitat Autònoma de Barcelona

References


Reis RJ, Utzet M, La Rocca PF, Nedel FB, Martin M, Navarro A. Previous sick leaves as predictor of subsequent ones. Int Arch Occup Environ Health 2011 Jun;84(5):491-499.


See Also

survsim-package, accum, mult.ev.sim, simple.surv.sim, crisk.sim

Examples

```r
### A cohort with 500 subjects, with a maximum follow-up time of 1825 days and
### just a covariate, following a Bernoulli distribution, and a corresponding
### beta of -0.4, -0.5, -0.6 and -0.7 for each episode. Note that random effect is
### the same for all events.

sim.data <- rec.ev.sim(n=500, foltime=1825, dist.ev=c('lnorm','llogistic',
                                     'weibull','weibull'),anc.ev=c(1.498, 0.924, 0.923, 1.051),beta0.ev=c(7.195,
                                     6.583, 6.678, 6.430),anc.cens=c(1.272, 1.218, 1.341, 1.484),
                                     beta0.cens=c(7.315, 6.975, 6.712, 6.399), z=list(c("unif", 0.8,1.2)),
                                     beta=list(c(-0.4,-0.5,-0.6,-0.7)), x=list(c("bern", 0.5)),
                                     lambda=c(2.18,2.33,2.40,3.46), priskb=0.5, max.old=730)

summary(sim.data)
```

Description

Simulation of cohorts in a context of standard survival analysis including several covariates and individual heterogeneity.
simple.surv.sim

Usage

simple.surv.sim(n, foltime, dist.ev, anc.ev, beta0.ev, dist.cens="weibull", anc.cens, beta0.cens, z=\text{NULL}, beta=\text{NA}, x=\text{NA})

Arguments

\begin{itemize}
\item \textbf{n} \quad \text{integer value indicating the desired size of the cohort to be simulated.}
\item \textbf{foltime} \quad \text{real number that indicates the maximum time of follow-up of the simulated cohort.}
\item \textbf{dist.ev} \quad \text{time to event distributions, with possible values weibull for the Weibull distribution, lnorm for the log-normal distribution and llogistic for the log-logistic distribution.}
\item \textbf{anc.ev} \quad \text{ancillary parameter for the time to event distribution.}
\item \textbf{beta0.ev} \quad \beta_0 \text{ parameter for the time to event distribution.}
\item \textbf{dist.cens} \quad \text{string indicating the time to censoring distributions, with possible values weibull for the Weibull distribution, lnorm for the log-normal distribution, llogistic for the log-logistic distribution and unif for the uniform distribution. If no distribution is introduced, the time to censoring is assumed to follow a Weibull distribution.}
\item \textbf{anc.cens} \quad \text{real number containing the ancillary parameter for the time to censoring distribution or the maximum in case of uniform distributed time to censoring.}
\item \textbf{beta0.cens} \quad \text{real number containing the } \beta_0 \text{ parameter for the time to censoring distribution or the minimum in case of uniform distributed time to censoring.}
\item \textbf{z} \quad \text{vector with three elements that contains information relative to a random effect used in order to introduce individual heterogeneity. The first element indicates the distribution: } \text{"unif" states for a uniform distribution, } \text{"gamma" states for a gamma distribution, } \text{"exp" states for an exponential distribution, } \text{"weibull" states for a Weibull distribution and } \text{"invgauss" states for an inverse gaussian distribution. The second and third elements indicates the minimum and maximum in the case of a uniform distribution (both must be positive) and the parameters in the case of the rest of distributions. Notice that that just one parameter is needed in the case of exponential distribution. Its default value is } \text{NULL}, \text{ indicating that no individual heterogeneity is introduced.}
\item \textbf{beta} \quad \text{list of elements indicating the effect of the corresponding covariate. The number of vectors in } \text{beta} \text{ must match the number of covariates. Its default value is } \text{NA}, \text{ indicating that no covariates are included.}
\item \textbf{x} \quad \text{list of vectors indicating the distribution and parameters of any covariate that the user needs to introduce in the simulated cohort. The possible distributions are } \text{"normal" for normal distribution, } \text{"unif" for uniform distribution and } \text{"bern" for Bernoulli distribution. Its default value is } \text{NA}, \text{ indicating that no covariates are included. The number of vectors in } x \text{ must match the number of vectors in } \text{beta}. \text{ Each vector in } x \text{ must contain the name of the distribution and the parameter(s), which are: the probability of success in the case of a Bernoulli distribution; the mean and the variance in the case of a normal distribution; and the minimum and maximum in the case of a uniform distribution.}
\end{itemize}
simple.surv.sim

Value

An object of class `simple.surv.sim`. It is a data frame containing the events suffered by the corresponding subjects. The columns of this data frame are detailed below:

- `nid`: an integer number that identifies the subject.
- `status`: logical value indicating if the corresponding event has been suffered or not.
- `start`: time at which the follow-up time begins for each event.
- `stop`: time at which the follow-up time ends for each event.
- `z`: Individual heterogeneity generated according to the specified distribution.
- `x`: value of each covariate randomly generated for each subject in the cohort.

Author(s)

David Moriña, Centre for Research in Environmental Epidemiology (CREAL) and Albert Navarro, Universitat Autònoma de Barcelona

References


See Also

`survsim-package, accum, rec.ev.sim, mult.ev.sim, crisk.sim`

Examples

```r
### A cohort with 1000 subjects, with a maximum follow-up time of 3600 days and two
### covariates, following a Bernoulli and uniform distribution respectively, and a
### corresponding beta of -0.4 for the first covariate and a corresponding beta of 0
### for the second covariate. Notice that the time to censorship is assumed to
### follow a Weibull distribution, as no other distribution is stated.

sim.data <- simple.surv.sim(n=1000, foltime=3600, dist.ev=c('llogistic'),
  anc.ev=c(0.69978200185280),beta0.ev=c(5.842985525742252),
  cens=1.17783687569519, beta0.cens=7.39773677281100,
  z=list(c("unif", 0.8, 1.2)), beta=list(c(-0.4), c(0)),
  x=list(c("bern", 0.5), c("unif", 0.7, 1.3)))

summary(sim.data)
```
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