# R documentation

of 'Ch1-PwrGSD.Rd' etc.

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PwrGSD

Calculate Power in a Group Sequential Design

## Description

Derives power in a two arm clinical trial under a group sequential design. Allows for arbitrary number of interim analyses, arbitrary specification of arm-0/arm-1 time to event distributions (via survival or hazard), arm-0/arm-1 censoring distribution, provisions for two types of continuous time non-compliance according to arm-0/arm-1 rate followed by switch to new hazard rate. Allows for analyses using (I) weighted log-rank statistic, with

weighting function (a) a member of the Flemming-Harrington G-Rho class, or (b) a stopped version thereof, or (c) the ramp-plateau deterministic weights, or (II) the integrated survival distance (currently under method=="S" without futility only). Stopping boundaries are computed via the Lan-Demets method, Haybittle method, or converted from the stochastic curtailment procedure. The Lan-Demets boundaries can be constructed usign either O'Brien-Flemming, Pocock or Wang-Tsiatis power alpha-spending. The C kernel is readily extensible, and further options will become available in the near future.

#### Usage

```
PwrGSD(EfficacyBoundary = LanDemets(alpha = 0.05, spending = ObrienFleming),
    FutilityBoundary = LanDemets(alpha = 0.1, spending = ObrienFleming),
    sided = c("2", ">", "<"), method = c("S", "A"), accru, accrat,</pre>
    tlook, tcut0 = NULL, h0 = NULL, s0 = NULL, tcut1 = NULL,
    rhaz = NULL, h1 = NULL, s1 = NULL, tcutc0 = NULL, hc0 = NULL,
    sc0 = NULL, tcutc1 = NULL, hc1 = NULL, sc1 = NULL, tcutd0A = NULL,
    hdOA = NULL, sdOA = NULL, tcutdOB = NULL, hdOB = NULL, sdOB = NULL,
    tcutd1A = NULL, hd1A = NULL, sd1A = NULL, tcutd1B = NULL,
    hd1B = NULL, sd1B = NULL, tcutx0A = NULL, hx0A = NULL, sx0A = NULL,
    tcutxOB = NULL, hxOB = NULL, sxOB = NULL, tcutx1A = NULL,
    hx1A = NULL, sx1A = NULL, tcutx1B = NULL, hx1B = NULL, sx1B = NULL,
    noncompliance = c("none", "crossover", "mixed", "user"),
    gradual = FALSE, WtFun = c("FH", "SFH", "Ramp"), ppar = cbind(c(0, 0)),
    Spend.Info = c("Variance", "Events", "Hybrid(k)", "Calendar"), RR.Futility = NULL,
    qProp.one.or.Q = c("one", "Q"), Nsim = NULL, detail = FALSE, StatType = c("WLR",
        "ISD"))
```

#### Arguments

#### EfficacyBoundary

This specifies the method used to construct the efficacy boundary. The available choices are

- (i) Lan-Demets(alpha=<total type I error>, spending=<spending function>). The Lan-Demets method is based upon a error probability spending approach. The spending function can be set to ObrienFleming, Pocock, or Power(rho), where rho is the the power argument for the power spending function: rho=3 is roughly equivalent to the O'Brien-Fleming spending function and smaller powers result in a less conservative spending function.
- (ii) Haybittle(alpha=<total type I error>, b.Haybittle=<user specified boundary point>). The Haybittle approach is the simplest, which sets the boundary points equal to b.Haybittle, a user specified value (try 3) for all analyses except the last, which is calculated so as to result in the total type I error, set with the argument alpha.
- (iii) SC(alpha=<total type I error>, crit=<threshold for conditional type I error for efficacy stopping>). The stochastic curtailment method is based upon the conditional probability of type I error given the current value of the statistic. Under this method, a sequence of boundary points on the standard normal scale (as are boundary points under all other methods) is calculated so that the total probability of type I error, alpha, is maintained. This is done by considering the joint probabilities of continuing to the current analysis and then exceeding the threshold at the

current analysis. A good value for the threshold value for the conditional type I error, crit is 0.90 or greater.

(iv) User supplied boundary points in the form c(b1, b2, b3, ..., b\_m), where m is the number of looks.

#### FutilityBoundary

This specifies the method used to construct the futility boundary. The available choices are

- (i) Lan-Demets(alpha=<total type II error>, spending=<spending function>). The Lan-Demets method is based upon a error probability spending approach. The spending function can be set to ObrienFleming, Pocock, or Power(rho), where rho is the the power argument for the power spending function: rho=3 is roughly equivalent to the O'Brien-Fleming spending function and smaller powers result in a less conservative spending function.
- (ii) Haybittle(alpha=<total type I error>, b.Haybittle=<user specified boundary point>). The Haybittle approach is the simplest, which sets the boundary points equal to b.Haybittle, a user specified value (try 3) for all analyses except the last, which is calculated so as to result in the total type II error, set with the argument alpha.
- (iii) SC(alpha=<total type II error>, crit=<threshold for conditional type II error for futility stopping>, drift.end=projected drift at end of trial>). The stochastic curtailment method is based upon the conditional probability of type II error given the current value of the statistic. Under this method, a sequence of boundary points on the standard normal scale (as are boundary points under all other methods) is calculated so that the total probability of type II error, alpha, is maintained. This is done by considering the joint probabilities of continuing to the current analysis and then exceeding the threshold at the current analysis. A good value for the threshold value for the conditional type I error, crit is 0.90 or greater.
- (iv) User supplied boundary points in the form c(b1, b2, b3, ..., b\_m), where m is the number of looks.

If two-sided tests of H0, set to "2" (quoted). If one-sided test of H0, set to ">" for upper tail, "<" for lower tail. If method=="S" then this must be of the same length as StatType because the interpretation of sided is different depending upon whether StatType=="WLR" (negative is benefit) or StatType=="ISD" (positive is benefit)

method Determines how to calculate the power. Set to "A" (Asymptotic method, the default) or "S" (Simulation method)

accru The upper endpoint of the accrual period beginning with time 0.

accrat The rate of accrual per unit of time.

s0

tlook The times of planned interim analyses.

tcut0 Left hand endpoints for intervals upon which the arm-0 specific mortality is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity.

A vector of the same length as tcut0 which specifies the piecewise constant arm-0 mortality rate.

Alternatively, the arm-0 mortality distribution can be supplied via this argument, in terms of of the corresponding survival function values at

the times given in the vector tcut0. If s0 is supplied, then h0is derived internally, assuming the piecewise exponential distribution. If you specify s0, the first element must be 1, and correspondingly, the first component of tcut0 will be the lower support point of the distribution. You must supply either h0 or s0 but not both.

tcut1

Left hand endpoints for intervals upon which the arm-1 specific mortality is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity.

rhaz

A vector of piecewise constant arm-1 versus arm-0 mortality rate ratios. If tcut1 and tcut0 are not identical, then tcut1, h0, and rhaz are internally rederived at the union of the sequences tcut0 and tcut1. In all cases the arm-1 mortality rate is then derived at the time cutpoints tcut1 as rhaz × h0.

h1

Alternatively, the arm-1 mortality distribution can be supplied via this argument by specifying the piecewise constant arm-1 mortality rate. See the comments above.

s1

Alternatively, the arm-1 mortality distribution can be supplied via this argument, in terms of of the corresponding survival function values at the times given in the vector tcut1. Comments regarding s0 above apply here as well. You must supply exactly one of the following: h1, rhaz, or s1.

tcutc0

Left hand endpoints for intervals upon which the arm-0 specific censoring distribution hazard function is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity.

hc0

A vector of the same length as tcutc0 which specifies the arm-0 censoring distribution in terms of a piecewise constant hazard function.

sc0

Alternatively, the arm-0 censoring distribution can be supplied via this argument, in terms of of the corresponding survival function values at the times given in the vector tcutc0. See comments above. You must supply either hc0 or sc0 but not both.

tcutc1

Left hand endpoints for intervals upon which the arm-1 specific censoring distribution hazard function is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity.

hc1

A vector of the same length as tcutc1 which specifies the arm-1 censoring distribution in terms of a piecewise constant hazard function.

sc1

Alternatively, the arm-1 censoring distribution can be supplied via this argument, in terms of of the corresponding survival function values at the times given in the vector tcutc1. See comments above. You must supply either hc1 or sc1 but not both.

noncompliance

(i) Seting noncompliance to "none" for no non-compliance will automatically set the non-compliance arguments, below, to appropriate values for no compliance. This requires no additional user specification of noncompliance parameters. (ii) Setting noncompliance to "crossover" will automatically set crossover values in the arm 0/1 specific post-cause-B-delay-mortality for cross-over, i.e. simple interchange of the arm 0 and arm 1 mortalities. The user is required to specify all parameters corresponding to the arm 0/1 specific cause-B-delay distributions. The cause-A-delay and post-cause-A-delay-mortality are automatically set so as not to influence the calculations. Setting noncompliance to "mixed"

will set the arm 0/1 specific post-cause-B-delay-mortality distributions for crossover as defined above. The user specifies the arm 0/1 specific cause-B-delay distribution as above, and in addition, all parameters related to the arm 0/1 specific cause-A-delay distributions and corresponding arm 0/1 specific post-cause-A-delay-mortality distributions. (iii) Setting noncompliance to "user" requires the user to specify all non-compliance distribution parameters.

tcutd0A

Left hand endpoints for intervals upon which the arm-0 specific *cause-A delay* distribution hazard function is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Required only when noncompliance is set to "mixed" or "user".

hdOA

A vector of the same length as tcutdOA containing peicewise constant hazard rates for the arm-0 cause-A delay distribution. Required only when noncompliance is set to "mixed" or "user".

sd0A

When required, the arm-0 *cause-A-delay* distribution is alternately specified via a survival function. A vector of the same length as tcutd0A.

tcutd0B

Left hand endpoints for intervals upon which the arm-0 specific *cause-B* delay distribution hazard function is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Always required when noncompliance is set to any value other than "none".

hd0B

A vector of the same length as tcutdOB containing peicewise constant hazard rates for the arm-0 *cause-B delay* distribution. Always required when noncompliance is set to any value other than "none".

sd0B

When required, the arm-0 cause-B-delay distribution is alternately specified via a survival function. A vector of the same length as tcutd0B.

tcutd1A

Left hand endpoints for intervals upon which the arm-1 specific *cause-A delay* distribution hazard function is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Required only when noncompliance is set to "mixed" or "user".

hd1A

A vector of the same length as tcutd1A containing peicewise constant hazard rates for the arm-1 *cause-A delay* distribution. Required only when noncompliance is set to "mixed" or "user".

sd1A

When required, the arm-1 cause-A-delay distribution is alternately specified via a survival function. A vector of the same length as tcutd1A.

tcutd1B

Left hand endpoints for intervals upon which the arm-1 specific *cause-B* delay distribution hazard function is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Always required when noncompliance is set to any value other than "none".

hd1B

A vector of the same length as tcutd1B containing peicewise constant hazard rates for the arm-1 *cause-B delay* distribution. Always required when noncompliance is set to any value other than "none".

sd1B

When required, the arm-1 *cause-A-delay* distribution is alternately specified via a survival function. A vector of the same length as tcutd1A.

tcutx0A

Left hand endpoints for intervals upon which the arm-0 specific *post-cause-A-delay-mortality* rate is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Required only when noncompliance is set to "mixed" or "user".

hxOA A vector of the same length as tcutxOA containing the arm-0 post-cause-A-delay mortality rates. Required only when noncompliance is set to "mixed" or "user".

When required, the arm-0 post-cause-A-delay mortality distribution is alternately specified via a survival function. A vector of the same length as tcutxOA.

Left hand endpoints for intervals upon which the arm-0 specific *post-cause-B-delay-mortality* rate is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Always required when noncompliance is set to any value other than "none".

A vector of the same length as tcutxOB containing the arm-0 post-cause-B-delay mortality rates. Always required when noncompliance is set to any value other than "none".

When required, the arm-0 post-cause-B-delay mortality distribution is alternately specified via a survival function. A vector of the same length as tcutx0B.

Left hand endpoints for intervals upon which the arm-1 specific *post-cause-A-delay-mortality* rate is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Required only when noncompliance is set to "mixed" or "user".

A vector of the same length as tcutx1A containing the arm-1 post-cause-A-delay mortality rates. Required only when noncompliance is set to "mixed" or "user".

When required, the arm-1 post-cause-A-delay mortality distribution is alternately specified via a survival function. A vector of the same length as tcutx1A.

Left hand endpoints for intervals upon which the arm-1 specific *post-cause-B-delay-mortality* rate is constant. The last given component is the left hand endpoint of the interval having right hand endpoint infinity. Always required when noncompliance is set to any value other than "none".

A vector of the same length as tcutx1B containing the arm-1 post-cause-B-delay mortality rates. Always required when noncompliance is set to any value other than "none".

When required, the arm-1 post-cause-B-delay mortality distribution is alternately specified via a survival function. A vector of the same length as tcutx1B.

Should the conversion to post-noncompliance mortality be gradual. Under the default behavior, <code>gradual=FALSE</code>, there is an immediate conversion to the post-noncompliance mortality rate function. If <code>gradual</code> is set to TRUE then this conversion is done "gradually". In truth, at the individual level what is done is that the new mortality rate function is a convex combination of the pre-noncompliance mortality and the post-noncompliance mortality, with the weighting in proportion to the time spent in compliance with the study arm protocal.

Specifies the name of a weighting function (of time) for assigning relative weights to events according to the times at which they occur. The default, "FH", a two parameter weight function, specifies the 'Fleming-Harrington' g-rho family of weighting functions defined as the pooled arm survival function (Kaplan-Meier estimate) raised to the g times its complement

sx0A

t.cutx0B

hx0B

sx0B

tcutx1A

hx1A

sx1A

tcutx1B

hx1B

sx1B

gradual

WtFun

raised to the rho. Note that g=rho=0 corresponds to the unweighted log-rank statistic. A second choice is the "SFH" function, (for 'Stopped Fleming-Harrington'), meaning that the "FH" weights are capped at their value at a user specified time, which has a total of 3 parameters. A third choice is Ramp(tcut). Under this choice, weights are assigned in a linearly manner from time 0 until a user specified cut-off time, tcut, after which events are weighted equally. It is possible to conduct computations on nstat candidate statistics within a single run. In this case, WtFun should be a character vector of length nstat having components set from among the available choices.

ppar

A vector containing all the weight function parameters, in the order determined by that of "WtFun". For example, if WtFun is set to c("FH", "SFH", "Ramp") then ppar should be a vector of length six, with the "FH" parameters in the first two elements, "SFH" parameters in the next 3 elements, and "Ramp" parameter in the last element.

RR.Futility

The relative risk corresponding to the alternative alternative hypothesis that is required in the construction of the futility boundary. Required if Boundary. Futility is set to a non-null value.

Spend.Info

When the test statistic is something other than the unweighted log-rank statistic, the variance information, i.e. the ratio of variance at interim analysis to variance at the end of trial, is something other than the ratio of events at interim analysis to the events at trial end. The problem is that in practice one doesn't necessarily have a good idea what the end of trial variance should be. In this case the user may wish to spend the type I and type II error probabilities according to a different time scale. Possible choices are "Variance", (default), which just uses the variance ratio scale, "Events", which uses the events ratio scale, "Hybrid(k)", which makes a linear transition from the "Variance" scale to the "Events" scale beginning with analysis number k. The last choice, "Calendar", uses the calendar time scale

qProp.one.or.Q

If a futility boundary is specified, what assumption should be made about the drift function (the mean value of the weighted log-rank statistic at analysis j normalized by the square root of the variance function at analysis k). In practice we don't presume to know the shape of the drift function. Set to "one" or "Q". The choice "one" results in a more conservative boundary.

Nsim

If you specify method=="S", then you must specify the number of simulations. 1000 should be sufficient.

detail

If you specify method=="S", and want to see the full level of detail regarding arguments returned from the C level code, specify detail==TRUE

StatType

If you specify method=="S", then the available choices are "WLR" (weighted log-rank) and "ISD" (integrated survival difference).

#### Value

Returns a value of class PwrGSD which has components listed below. Note that the print method will display a summary table of estimated powers and type I errors as a nstat by 2 matrix. The summary method returns the same object invisibly, but after computing the summary table mentioned above, and it is included in the returned value as a commponent TBL. See examples below.

dPower A length(tlook) by nstat matrix containing in each column, an increment in power that resulted at that analysis time for the given statistic.

dErrorI A length(tlook) by nstat matrix containing in each column, an increment in type I error that resulted at that analysis time for the given

statistic. Always sums to the total alpha specified in alphatot

detail A list with components equal to the arguments of the C-call, which cor-

respond in a natural way to the arguments specified in the R call, along with the computed results in palphaOvec, palphaIvec, inffrac, and mu. The first two are identical to dErrorI and dPower, explained above. The last two are length(tlook) by nstat matrices. For each statistic specified in par, the corresponding columns of pinffrac and mu contain the

information fraction and drift at each of the analysis times.

call the call

## Author(s)

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#### References

Gu, M.-G. and Lai, T.-L. "Determination of power and sample size in the design of clinical trials with failure-time endpoints and interim analyses." Controlled Clinical Trials 20 (5): 423-438, 1999

Izmirlian, G. "The PwrGSD package." NCI Div. of Cancer Prevention Technical Report. 2004

Jennison, C. and Turnbull, B.W. (1999) Group Sequential Methods: Applications to Clinical Trials Chapman & Hall/Crc, Boca Raton FL

Proschan, M.A., Lan, K.K.G., Wittes, J.T. (2006), corr 2nd printing (2008) Statistical Monitoring of Clinical Trials A Unified Approach Springer Verlag, New York

#### See Also

cpd.PwrGSD

```
library(PwrGSD)
```

cpd.PwrGSD

Create a skeleton compound PwrGSD object

#### Description

Given a user defined indexing dataframe as its only argument, creates a skeleton compound PwrGSD object having a component Elements, a list of PwrGSD objects, of length equal to the number of rows in the indexing dataframe

#### Usage

```
cpd.PwrGSD(descr)
```

#### Arguments

descr

A dataframe of a number of rows equal to the length of the resulting list, Elements, of PwrGSD objects. The user defines the mapping between rows of descr and components of Elements and uses it to set up a loop over scenarios. There are several S3 classes and methods for example plot.cpd.PwrGSD, which exploit this mapping between characteristics of a run and the rows of desr for subsetting and constructing conditioned plots. See the example below.

#### Value

An object of class cpd.PwrGSD containing elements:

date the POSIX date that the object was created-its quite useful

Elements a list of length equal to the number of rows of descr which will later

contain objects of class PwrGSD

descr a copy of the indexing dataframe argument for use in navigating the com-

pound object in subsequent calls to other functions such as the related

plot method, and the subset extractor, Elements

#### Note

A cpd.PwrGSD object essentially a list of PwrGSD objects that a user may set up in order to investigate the space of possible trial scenarios, test statistics, and boundary construction options. One could store a list of results without appealing at all to these internal indexing capabilities. The advantage of setting up a cpd.PwrGSD object is the nice summarization functionality provided, for example the plot method for the cpd.PwrGSD class.

The key ingredient to (i) the construction of the empty object, (ii) and summarizing the results in tabular or plotted form via its manipulation in subsequent function calls, is the indexing dataset, descr (for description). The correspondence between rows of descr and

elements in the list of PwrGSD objects is purposely left very loose. In the example outlined below, the user creates a "base case" call to PwrGSD and then decides which quantities in this "base case" call to vary in order to navigate the space of possible trial scenarios, monitoring statistics and boundary construction methods. Next, for each one of these settings being varied, a variable with levels that determine each possible setting is created. The dataset descr is created with one line corresponding to each combination of the selection variables so created. In order to ensure that there is 1-1 correspondence between the order of the rows in descr and the order in the list Elements of PwrGSD objects, the user carries out the computation in a loop over rows of descr in which the values of the selection variables in each given row of descr are used to create the corresponding component of Elements via an update the "base case" call.

## Author(s)

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#### See Also

Elements, plot.cpd.PwrGSD and Power

```
## don't worry--these examples are guaranteed to work,
## its just inconvenient for the package checker
## Not run:
  library(PwrGSD)
## In order to set up a compound object of class `cpd.PwrGSD'
## we first construct a base case: a two arm trial randomized in just
## under eight years with a maximum of 20 years of follow-up.
## We compute power at a specific alternative, `rhaz', under
## an interim analysis plan with roughly one annual analysis, some
## crossover between intervention and control arms, with Efficacy
## and futility boundaries constructed via the Lan-Demets procedure
## with O'Brien-Fleming spending on the hybrid scale. Investigate
## the behavior of three weighted log-rank statistics.
test.example <-
  PwrGSD(EfficacyBoundary = LanDemets(alpha = 0.05, spending = ObrienFleming),
         FutilityBoundary = LanDemets(alpha = 0.1, spending = ObrienFleming),
         RR.Futility = 0.82, sided="<",method="A",accru=7.73, accrat=9818.65,
         tlook =c(7.14, 8.14, 9.14, 10.14, 10.64, 11.15, 12.14, 13.14,
                  14.14, 15.14, 16.14, 17.14, 18.14, 19.14, 20.14),
         tcut0 =0:19, h0 =c(rep(3.73e-04, 2), rep(7.45e-04, 3),
                            rep(1.49e-03, 15)),
         tcut1 =0:19, rhaz =c(1, 0.9125, 0.8688, 0.7814, 0.6941,
                              0.6943, 0.6072, 0.5202, 0.4332, 0.6520,
                              0.6524, 0.6527, 0.6530, 0.6534, 0.6537,
                              0.6541, 0.6544, 0.6547, 0.6551, 0.6554),
         tcutc0 = 0:19, hc0 = c(rep(1.05e-02, 2), rep(2.09e-02, 3),
                              rep(4.19e-02, 15)),
         tcutc1 =0:19, hc1 =c(rep(1.05e-02, 2), rep(2.09e-02, 3),
                              rep(4.19e-02, 15)),
         tcutd0B = c(0, 13), hd0B = c(0.04777, 0),
         tcutd1B =0:6, hd1B =c(0.1109, 0.1381, 0.1485, 0.1637, 0.2446,
                               0.2497, 0),
```

```
noncompliance =crossover, gradual =TRUE,
         WtFun =c("FH", "SFH", "Ramp"),
         ppar =c(0, 1, 0, 1, 10, 10))
## we will construct a variety of alternate hypotheses relative to the
## base case specified above
  rhaz <-
    c(1, 0.9125, 0.8688, 0.7814, 0.6941, 0.6943, 0.6072, 0.5202, 0.4332,
    0.652, 0.6524, 0.6527, 0.653, 0.6534, 0.6537, 0.6541, 0.6544,
    0.6547, 0.6551, 0.6554)
  max.effect <- 0.80 + 0.05*(0:8)
  n.me <- length(max.effect)</pre>
## we will also vary extent of censoring relative to the base case
## specified above
  hc \leftarrow c(rep(0.0105, 2), rep(0.0209, 3), rep(0.0419, 15))
  cens.amt <- 0.75 + 0.25*(0:2)
  n.ca <- length(cens.amt)</pre>
## we may also wish to compare the Lan-Demets/0'Brien-Fleming efficacy
## boundary with a Pocock efficacy boundary
  Eff.bound.choice <- 1:2
  ebc.nms <- c("LanDemets(alpha=0.05, spending=ObrienFleming)",</pre>
               "SC(alpha=0.05, crit=0.90)")
 n.ec <- length(Eff.bound.choice)</pre>
## The following line creates the indexing dataframe, `descr', with one
## line for each possible combination of the selection variables we've
## created.
  descr <- as.data.frame(</pre>
              cbind(Eff.bound.choice=rep(Eff.bound.choice, each=n.ca*n.me),
                    cens.amt=rep(rep(cens.amt, each=n.me), n.ec),
                    max.effect=rep(max.effect, n.ec*n.ca)))
  descr$Eff.bound.choice <- ebc.nms[descr$Eff.bound.choice]</pre>
## Now descr contains one row for each combination of the levels of
## the user defined selection variables, `Eff.bound.choice',
## `max.effect' and `cens.amt'. Keep in mind that the names and number
## of these variables is arbitrary. Next we create a skeleton
## `cpd.PwrGSD' object with a call to the function `cpd.PwrGSD' with
## argument `descr'
  test.example.set <- cpd.PwrGSD(descr)</pre>
## Now, the newly created object, of class `cpd.PwrGSD', contains
## an element 'descr', a component 'date', the date created
## and a component `Elements', an empty list of length equal
## to the number of rows in `descr'. Next we do the computation in
## a loop over the rows of `descr'.
```

```
n.descr <- nrow(descr)</pre>
for(k in 1:n.descr){
  ## First, we copy the original call to the current call,
  ## `Elements[[k]]$call'
 test.example.set$Elements[[k]]$call <- test.example$call</pre>
  ## Use the efficacy boundary choice in the kth row of `descr'
  ## to set the efficacy boundary choice in the current call
  test.example.set$Elements[[k]]$call$EfficacyBoundary <-</pre>
  parse(text=as.character(descr[k,"Eff.bound.choice"]))[[1]]
  ## Derive the `rhaz' defined by the selection variable "max.effect"
  ## in the kth row of `descr' and use this to set the `rhaz'
  ## components of the current call
  test.example.set$Elements[[k]]$call$rhaz <-</pre>
                          exp(descr[k,"max.effect"] * log(rhaz))
  ## Derive the censoring components from the selection variable
  ## "cens.amt" in the kth row of `descr' and place that result
  ## into the current call
  test.example.set$Elements[[k]]$call$hc0 <-</pre>
  test.example.set$Elements[[k]]$call$hc1 <-</pre>
                              exp(descr[k, "cens.amt"] * log(hc))
 ## Now the current call corresponds exactly to the selection
  ## variable values in row `k' of `descr'. The computation is
  ## done by calling `update'
  test.example.set$Elements[[k]] <-</pre>
                        update(test.example.set$Elements[[k]])
  cat(k/n.descr, "\r")
## We can create a new `cpd.PwrGSD' object by subsetting on
## the selection variables in `descr':
Elements(test.example.set,
         subset=(substring(Eff.bound.choice, 1,9)=="LanDemets" &
                          max.effect >= 1))
## or we can plot the results -- see the help under `plot.cpd.PwrGSD'
plot(test.example.set, formula = ~ max.effect | stat * cens.amt,
     subset=(substring(Eff.bound.choice, 1,9)=="LanDemets"))
plot(test.example.set, formula = ~ max.effect | stat * cens.amt,
     subset=(substring(Eff.bound.choice, 1,2)=="SC"))
## Notice the appearance of the selection variable `stat' which was
## not defined in the dataset `descr'.
```

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Power

Extract the Power results

#### Description

Extract the Power results from a compound object into a Stacked Dataframe

#### Usage

```
Power(object, subset)
```

#### Arguments

object an object of class cpd.PwrGSD

subset you may extract a subset via a logical expression in the variables of the

index dataframe, descr

## Value

an object of class cpd.PwrGSD. See help on that topic for details.

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

#### See Also

```
cpd.PwrGSD and PwrGSD
```

```
## See the `cpd.PwrGSD' example
```

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Elements

Create a subset of a "cpd.PwrGSD" object

## Description

Create a subset of a cpd.PwrGSD object

#### Usage

```
Elements(object, subset, na.action = na.pass)
```

## Arguments

object an object of class cpd.PwrGSD

subset you may extract a subset via a logical expression in the variables of the

index dataframe, descr

na.action a method for handling NA values in the variables in the subset expression.

#### Value

an object of class cpd.PwrGSD. See help on that topic for details.

#### Author(s)

Grant Izmirlian <izmirlian@nih.gov>

#### See Also

```
cpd.PwrGSD and PwrGSD
```

## Examples

```
## See the `cpd.PwrGSD' example
```

plot.cpd.PwrGSD

Plot Method for cpd.PwrGSD objects

## Description

Creates a trellis plot of type II error probability and power at each interim analysis, stacked, versus an effect size variable, conditioned upon levels of up to two factors.

## Usage

```
## S3 method for class 'cpd.PwrGSD':
plot(x, formula, subset, na.action,...)
```

#### Arguments

x an object of class cpd.PwrGSD

formula a one sided formula of the form ~ effect | f1 or ~ effect | f1 \*

f2 where effect, f1, and f2 are variables in the indexing dataframe descr, or the special variable stat which may be used when there are multiple test statistics per component of Elements. See the example in

the documentation for cpd.PwrGSD

subset the plot can be applied to a subset of rows of descr via a logical expres-

sion on its variables in combination with the special variable, stat when

applicable.

na.action a na.action method for handling NA values

... other parameters to pass to the R function coplot usually not necessary

#### Value

Returns the object, x, invisibly

#### Note

This processes the cpd.PwrGSD object into a dataframe, stacked on interim looks and then passes the results to the R function coplot

#### Author(s)

Abovementioned cpd.PwrGSD processing done by Grant Izmirlian <izmirlian@nih.gov>

#### References

Chambers, J. M. (1992) Data for models. Chapter 3 of Statistical Models in S eds J. M. Chambers and T. J. Hastie, Wadsworth & Brooks/Cole.

Cleveland, W. S. (1993) Visualizing Data. New Jersey: Summit Press.

## See Also

cpd.PwrGSD Power and Elements

#### Examples

## See the example in the 'cpd.PwrGSD' documentation

GrpSeqBnds

Computes efficacy and futility boundaries

#### Description

This computes efficacy and futility boundaries for interim analysis and sequential designs. Two sided symmetric efficacy boundaries can be computed by specifying half of the intended total type I error probability in the argument, Alpha.Efficacy. Otherwise, especially in the case of efficacy and futility bounds only one sided boundaries are currently computed. The computation allows for two different time scales—one must be the variance ratio, and the second can be a user chosen increasing scale beginning with 0 that takes the value 1 at the conclusion of the trial.

#### Usage

#### Arguments

#### EfficacyBoundary

This specifies the method used to construct the efficacy boundary. The available choices are

- (i) Lan-Demets(alpha=<total type I error>, spending=<spending function>). The Lan-Demets method is based upon a error probability spending approach. The spending function can be set to ObrienFleming, Pocock, or Power(rho), where rho is the the power argument for the power spending function: rho=3 is roughly equivalent to the O'Brien-Fleming spending function and smaller powers result in a less conservative spending function.
- (ii) Haybittle(alpha=<total type I error>, b.Haybittle=<user specified boundary point>). The Haybittle approach is the simplest, which sets the boundary points equal to b.Haybittle, a user specified value (try 3) for all analyses except the last, which is calculated so as to result in the total type I error, set with the argument alpha.
- (iii) SC(alpha=<total type I error>, crit=<threshold for conditional type I error for efficacy stopping>). The stochastic curtailment method is based upon the conditional probability of type I error given the current value of the statistic. Under this method, a sequence of boundary points on the standard normal scale (as are boundary points under all other methods) is calculated so that the total probability of type I error, alpha, is maintained. This is done by considering the joint probabilities of continuing to the current analysis and then exceeding the threshold at the current analysis. A good value for the threshold value for the conditional type I error, crit is 0.90 or greater.
- (iv) User supplied boundary points in the form c(b1, b2, b3, ..., b\_m), where m is the number of looks.

## FutilityBoundary

This specifies the method used to construct the futility boundary. The available choices are

- (i) Lan-Demets(alpha=<total type II error>, spending=<spending function>). The Lan-Demets method is based upon a error probability spending approach. The spending function can be set to ObrienFleming, Pocock, or Power(rho), where rho is the the power argument for the power spending function: rho=3 is roughly equivalent to the O'Brien-Fleming spending function and smaller powers result in a less conservative spending function.
- (ii) Haybittle(alpha=<total type I error>, b.Haybittle=<user specified boundary point>). The Haybittle approach is the simplest, which sets the boundary points equal to b.Haybittle, a user specified value (try 3) for all analyses except the last, which is calculated so as to result in the total type II error, set with the argument alpha.
- (iii) SC(alpha=<total type II error>, crit=<threshold for conditional type II error for futility stopping>, drift.end=cprojected drift

at end of trial>). The stochastic curtailment method is based upon the conditional probability of type II error given the current value of the statistic. Under this method, a sequence of boundary points on the standard normal scale (as are boundary points under all other methods) is calculated so that the total probability of type II error, alpha, is maintained. This is done by considering the joint probabilities of continuing to the current analysis and then exceeding the threshold at the current analysis. A good value for the threshold value for the conditional type I error, crit is 0.90 or greater.

(iv) User supplied boundary points in the form c(b1, b2, b3,..., b\_m),

where m is the number of looks.

frac The variance ratio. If the end of trial variance is unknown then normalize

all previous variances by the current variance. In this case you must specify a second scale that is monotone increasing from 0 to 1 at the end

of the trial. Required.

 ${\tt frac.ii} \qquad \qquad {\tt The \ second \ information \ scale \ that \ is \ used \ for \ type \ I \ and \ type \ II \ error}$ 

probability spending. Optional (see above)

drift The drift function of the underlying brownian motion, which is the ex-

pected value under the design alternative of the un-normalized weighted log-rank statistic, then normalized to have variance one when the variance

ratio equals 1. See the examples below.

#### Value

An object of class boundaries with components: "table" "frac" "frac.ii" "drift" "call"

The call that produced the returned results.

frac The vector of variance ratios.

frac.ii The vector of information ratios for type I and type II error probabil-

ity spending, which differs from the above if the user sets the argument

frac.ii to a second scale as mentioned above.

drift The drift vector that is required as an argument when futility boundaries

are calculated.

table A matrix with components

frac The information ratio for type I and type II error probability spending.

b.f The calculated futility boundary (if requested).

alpha.f The type II error probability spent at that analysis (if doing futility

bounds).

cum-alpha.f Cumulative sum of alpha.f (if doing futility bounds).

b.e The calculated efficacy boundary.

alpha.e The type I error probability spent at that analysis.

cum-alpha.e Cumulative sum of alpha.e.

#### Author(s)

Grant Izmirlian <izmirlian@nih.gov>

#### References

Gu, M.-G. and Lai, T.-L. "Determination of power and sample size in the design of clinical trials with failure-time endpoints and interim analyses." Controlled Clinical Trials 20 (5): 423-438, 1999

Izmirlian, G. "The PwrGSD package." NCI Div. of Cancer Prevention Technical Report. 2004

Jennison, C. and Turnbull, B.W. (1999) Group Sequential Methods: Applications to Clinical Trials Chapman & Hall/Crc, Boca Raton FL

Proschan, M.A., Lan, K.K.G., Wittes, J.T. (2006), corr 2nd printing (2008) Statistical Monitoring of Clinical Trials A Unified Approach Springer Verlag, New York

#### See Also

PwrGSD

#### Examples

```
## NOTE: In an unweighted analysis, the variance ratios and event ratios
## are the same, whereas in a weighted analysis, they are quite different.
## For example, in a trial with 7 or so years of accrual and maximum follow-up of 20 years
## using the stopped Fleming-Harrington weights, `WtFun' = "SFH", with paramaters
## `ppar' = c(0, 1, 10) we might get the following vector of variance ratios:
        <- c(0.006995655, 0.01444565, 0.02682463, 0.04641363, 0.0585665,</pre>
frac
             0.07614902, 0.1135391, 0.168252, 0.2336901, 0.3186155, 0.4164776,
             0.5352199, 0.670739, 0.8246061, 1)
## and the following vector of event ratios:
frac.ii <-c(0.1494354, 0.1972965, 0.2625075, 0.3274323, 0.3519184, 0.40231,
             0.4673037, 0.5579035, 0.6080742, 0.6982293, 0.7671917, 0.8195019,
             0.9045182, 0.9515884, 1)
## and the following drift under a given alternative hypothesis
           c(0.06214444, 0.1061856, 0.1731267, 0.2641265, 0.3105231, 0.3836636,
drift <-
             0.5117394, 0.6918584, 0.8657705, 1.091984, 1.311094, 1.538582,
             1.818346, 2.081775, 2.345386)
## JUST ONE SIDED EFFICACY BOUNDARY
## In this call, we calculate a one sided efficacy boundary at each of 15 analyses
## which will occur at the given (known) variance ratios, and we use the variance
## ratio for type I error probability spending, with a total type I error probabilty
## of 0.05, using the Lan-Demets method with Obrien-Fleming spending (the default).
gsb.all.just.eff <- GrpSeqBnds(frac=frac,</pre>
                               EfficacyBoundary=LanDemets(alpha=0.05, spending=ObrienFleming))
## ONE SIDED EFFICACY AND FUTILTY BOUNDARIES
## In this call, we calculate a one sided efficacy boundary at each of 15 analyses
## which will occur at the given (known) variance ratios, and we use the variance
```

## ratio for type I and type II error probability spending, with a total type I error ## probability of 0.05 and a total type II error probability of 0.10, using the Lan-Demets

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```
## method with Obrien-Fleming spending (the default) for both efficacy and futilty.
gsb.all.eff.fut <- GrpSeqBnds(frac=frac,</pre>
                              EfficacyBoundary=LanDemets(alpha=0.05, spending=ObrienFleming),
                              FutilityBoundary=LanDemets(alpha=0.10, spending=ObrienFleming),
                              drift=drift)
## Now suppose that we are performing the 7th interim analysis. We don't know what the variance
## will be at the end of the trial, so we normalize variances of the current and previous
## statistics by the variance of the current statistic. This is equivalent to the following
## length 7 vector of variance ratios:
frac7 <- frac[1:7]/frac[7]</pre>
## To proceed under the "unknown variance at end of trial" case, we must use a second
## scale for spending type I and II error probabilty. Unlike the above scale
## which is renormalized at each analysis to have value 1 at the current analysis, the
## alpha spending scale must be monotone increasing and attain the value 1 only at the
## end of the trial. A natural choice is the event ratio, which is known in advance if
## the trial is run until a required number of events is obtained, a so called
## maximum information trial:
frac7.ii <- frac.ii[1:7]</pre>
## the first seven values of the drift function
drift7 <- drift[1:7]/frac[7]^0.5</pre>
gsb.1st7.eff.fut <- GrpSeqBnds(frac=frac7, frac.ii=frac7.ii,</pre>
                               EfficacyBoundary=LanDemets(alpha=0.05, spending=ObrienFleming),
                               FutilityBoundary=LanDemets(alpha=0.10, spending=ObrienFleming),
                               drift=drift7)
## Of course there are other options not covered in these examples but this should get you
## started
```

CondPower

Conditional type I and type II error probabilities given current value of the test statistic

#### Description

Computes conditional type I and type II error probabilities given current value of the test statistic for monitoring based upon stochastic curtailment. This is now obsolete and included in the functionality of "GrpSeqBnds" and is here for instructional purposes only.

#### Usage

```
CondPower(Z, frac, drift, drift.end, err.I, sided = 1)
```

#### Arguments

Z Current value of test statistic standardized to unit variance.

frac Current value of the information fraction (variance fraction).

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drift	Current value of the drift, i.e. the expected value of the test statistic
	normalized to have variance equal to the information fraction. Required
	if you want to compute conditional type II error, otherwise enter 0.
drift.end	Projected value of the drift at the end of the trial.
err.I	Overall (total) type I error probability
sided	Enter 1 or 2 for sided-ness of the test.

## Value

A named numeric vector containing the two components "Pr.cond.typeIerr" and "Pr.cond.typeIIerr"

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

#### References

A General Theory on Stochastic Curtailment for Censored Survival Data D. Y. Lin, Q. Yao, Zhiliang Ying Journal of the American Statistical Association, Vol. 94, No. 446 (Jun., 1999), pp. 510-521

#### See Also

GrpSeqBnds

#### Examples

## None as yet

SimGSB

Verifies the results of "GrpSeqBnds" via simulation

#### Description

Verifies the results of GrpSeqBnds via simulation

### Usage

```
SimGSB(object, nsim = 1e+05, ...)
```

## Arguments

object an object of class either boundaries or PwrGSD

nsim number of simulations to do

... if object is of class PwrGSD and there are more than one statistic under investigation than you may enceify an argument stat. The default value

investigation, then you may specify an argument  ${\tt stat}.$  The default value

is 1, meaning the first one.

#### Value

A tabulation of the results

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#### Author(s)

Grant Izmirlian <izmirlian@nih.gov

#### See Also

GrpSeqBnds

#### Examples

## none as yet

as.boundaries

 $Convert\ a\ "PwrGSD"\ object\ to\ a\ "boundaries"\ object$ 

#### Description

Convert a PwrGSD object to a boundaries object

## Usage

```
as.boundaries(object, ...)
```

## Arguments

object an object of class PwrGSD

if object is of class PwrGSD and there are more than one statistic under . . . investigation, then you may specify an argument stat. The default value

is 1, meaning the first one.

## Value

an object of class boundaries. See the documentation for GrpSeqBnds

## Author(s)

Grant Izmirlian <izmirlian@nih.gov

#### See Also

 ${\tt GrpSeqBnds}$ 

```
## none as yet
```

22 wtdlogrank

wtdlogrank Weighted log-rank test
-----------------------------------

#### Description

Computes a two sample weighted log-rank statistic with events weighted according to one of the available weighting function choices

#### Usage

```
wtdlogrank(formula = formula(data), data = parent.frame(), WtFun = c("FH", "SFH", "Ramp") param = c(0, 0), sided = c(2, 1), subset, na.action, w = FALSE)
```

#### Arguments

formula a formula of the form Surv(Time, Event) ~ arm where arm is a dichoto-

mous variable with values 0 and 1.

data a dataframe

WtFun a selection from the available list: "FH" (Fleming-Harrington), "SFH"

(stopped Fleming-Harrington) or "Ramp". See param in the following

line.

param Weight function parameters. Length and interpretation depends upon the

selected value of WtFun: If WtFun==\dQuote{FH} then param is a length 2 vector specifying the power of the pooled (across arms) kaplan meier estimate and its complement. If WtFun==\dQuote{SFH} then param is a length 3 vector with first two components as in the "FH" case, and third component the time (in the same units as the time to event) at which the "FH" weight function is capped off at its current value. If WtFun==\dQuote{SFH} then param is of length 1 specifying the time (same units as time to event) at which events begin to get equal weight. The "Ramp" weight function is a linearly increasing deterministic weight func-

tion which is capped off at 1 at the user specified time.

one or Two sided test? Set to 1 or 2

subset Analysis can be applied to a subset of the dataframe based upon a logical

expression in its variables

na.action Method for handling NA values in the covariate, arm

w currently no effect

#### Value

An object of class survtest containing components

pn sample size

wttyp internal representation of the WtFun argument
par internal representation of the param argument

time unique times of events accross all arms

nrisk number at risk accross all arms at each event time

nrisk1 Number at risk in the experimental arm at each event time

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nevent	Number of events across all arms at each event time
nevent1	Number of events in the experimental arm at each event time
wt	Values of the weight function at each event time
pntimes	Number of event times
stat	The un-normalized weighted log-rank statistic, i.e. the summed weighted observed minus expected differences at each event time
var	Variance estimate for the above
UQt	Cumulative sum of increments in the sum resulting in stat above
varQt	Cumulative sum of increments in the sum resulting in var above
var1t	Cumulative sum of increments in the sum resulting in the variance of an unweighted version of the statistic
pu0	person units of follow-up time in the control arm
pu1	person units of follow-up time in the intervention arm
n0	events in the control arm
n1	events in the intervention arm
n	sample size, same as pn
call	the call that created the object

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

#### References

Harrington, D. P. and Fleming, T. R. (1982). A class of rank test procedures for censored survival data. *Biometrika* **69**, 553-566.

#### See Also

 ${\tt IntSurvDiff}$ 

## Examples

```
library(PwrGSD)
data(lung)
fit.wlr <- wtdlogrank(Surv(time, I(status==2))~I(sex==2), data=lung, WtFun="SFH", param=c(0,1,300))</pre>
```

IntSurvDiff Weighted Integrated Survival function test

#### Description

Computes a two sample weighted integrated survival function log-rank statistic with events weighted according to one of the available weighting function choices

## Usage

```
 IntSurvDiff(formula = formula(data), data = parent.frame(), WtFun = c("FH", "SFH", "Ramp" param = c(0, 0), sided = c(2, 1), subset, na.action, w = FALSE)
```

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#### Arguments

formula a formula of the form Surv(Time, Event) ~ arm where arm is a dichoto-

mous variable with values 0 and 1.

data a dataframe

WtFun a selection from the available list: "FH" (Fleming-Harrington), "SFH"

(stopped Fleming-Harrington) or "Ramp". See param in the following

line.

param Weight function parameters. Length and interpretation depends upon the

selected value of WtFun: If WtFun==\dQuote{FH} then param is a length 2 vector specifying the power of the pooled (across arms) kaplan meier estimate and its complement. If WtFun==\dQuote{SFH} then param is a length 3 vector with first two components as in the "FH" case, and third component the time (in the same units as the time to event) at which the "FH" weight function is capped off at its current value. If WtFun==\dQuote{SFH} then param is of length 1 specifying the time (same units as time to event) at which events begin to get equal weight. The "Ramp" weight function is a linearly increasing deterministic weight func-

tion which is capped off at 1 at the user specified time.

sided One or Two sided test? Set to 1 or 2

subset Analysis can be applied to a subset of the dataframe based upon a logical

expression in its variables

na.action Method for handling NA values in the covariate, arm

w currently no effect

#### Value

An object of class survtest containing components

pn sample size

wttyp internal representation of the WtFun argument par internal representation of the param argument

time unique times of events across all arms

nrisk number at risk accross all arms at each event time

nrisk1 Number at risk in the experimental arm at each event time

nevent Number of events across all arms at each event time

nevent1 Number of events in the experimental arm at each event time

wt Values of the weight function at each event time

pntimes Number of event times

stat The un-normalized weighted log-rank statistic, i.e. the summed weighted

observed minus expected differences at each event time

var Variance estimate for the above

pu0 person units of follow-up time in the control armpu1 person units of follow-up time in the intervention arm

n0 events in the control armn1 events in the intervention arm

n sample size, same as pn

call the call that created the object

mysurvfit 25

#### Author(s)

Grant Izmirlian <izmirlian@nih.gov

#### References

Weiand S, Gail MH, James BR, James KL. (1989). A family of nonparametric statistics for comparing diagnostic makers with paired or unpaired data. *Biometrika* **76**, 585-592.

#### See Also

wtdlogrank

## Examples

```
library(PwrGSD)
data(lung)
fit.isd <- IntSurvDiff(Surv(time, I(status==2))~I(sex==2), data=lung, WtFun="SFH", param=c(0,1,300))</pre>
```

mysurvfit

My Survfit

#### Description

Computes numbers at risk, numbers of events at each unique event time within levels of a blocking factor

## Usage

```
mysurvfit(formula = formula(data), data = parent.frame(), subset, na.action = na.fail)
```

## Arguments

formula	Should be a formula of the form Surv(ti, ev) ~ block where block is the blocking factor. It need not be a factor per se but should have relatively few discrete levels. Sorry, no staggered entry allowed at present
data	a dataframe
subset	you can subset the analysis via logical expression in variables in the dataframe
na.action	pass a method for handling NA values in $block$ such as $na.omit$ , or $na.fail$

## Value

A dataframe of 2\*NLEV + 1 columns where NLEV is the number of levels of the factor block.

time	The sorted vector of unique event times from all blocks
nrisk1	The number at risk in block level 1 at each event time
nevent1	The number of events in block level 1 at each event time $$

. . .

nriskNLEV The number at risk in block level NLEV at each event time

neventNLEV The number of events in block level NLEV at each event time

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#### Author(s)

Grant Izmirlian <izmirlian@nih.gov>

#### Examples

```
library(PwrGSD)
data(lung)

fit.msf <- mysurvfit(Surv(time, I(status==2)) ~ sex, data=lung)

fit.msf
## Not run:
plot(fit.msf)

## End(Not run)</pre>
```

agghaz

Aggregated Hazard

## Description

Computes the MLE for the model that assumes piecewise constant hazards on intervals defined by a grid of points. One applications for example is to calculate monthly hazard rates given numbers of events, numbers at risk and event times reported to the day. Can also handle time to event data stratified on a blocking factor.

#### Usage

```
agghaz(t.agg, time, nrisk, nevent)
```

#### Arguments

t.agg	Vector defining intervals upon which the user wants constant hazard rates.
time	Event times, possibly stratified on a blocking factor into multiple columns, in units that occur in enough numbers per interval specified above. If there is just a single column then it must be in column form (see example below).
nrisk	Numbers at risk at specified event times
nevent	Numbers of events at specified event times

## Value

time.a	User supplied left-hand endpoints of intervals of hazard constancy
nrisk.a	Numbers at risk on specified intervals
nevent.a	Numbers of events on specified intervals

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

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#### Examples

```
library(PwrGSD)
data(lung)
fit.msf <- mysurvfit(Surv(time, I(status==2)) ~ sex, data=lung)

## A single stratum:
with(fit.msf$Table, agghaz(30, time, cbind(nrisk1), cbind(nevent1)))

## Multiple strata--pooled and group 1:
with(fit.msf$Table, agghaz(30, time, cbind(nrisk1+nrisk2,nrisk1), cbind(nevent1+nevent2,nevent1)))</pre>
```

mystack

Stack a dataset

#### Description

Given a dataframe containing one or more variables named with a common prefix, this function creates a stacked dataset with one set of observed values of the variables (in order of occurence) per line.

#### Usage

```
mystack(object, fu.vars, create.idvar = FALSE)
```

## Arguments

object a dataframe containing one or more variables named with a common prefix

fu.vars a list of the unique prefixes

create.idvar Do you want to add an ID variable with a common value given to all

records resulting from a given input record? Default is  ${\tt FALSE}$ 

#### Value

A stacked dataframe

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

```
## none as yet
```

28 CDFOR2LRR

CDFOR2LRR

 $Convert\ CDF\ Odds\ Ratio\ to\ Logged\ Relative\ Risks$ 

## Description

A concise (1-5 lines) description of what the function does.

## Usage

```
CDFOR2LRR(tcut, tmax, h0, CDFOR)
```

## Arguments

tcut	Describe tcut here
tmax	Describe tmax here
h0	Describe h0 here
CDFOR	Describe CDFOR here

## Details

If necessary, more details than the description above

## Value

Describe the value returned If it is a LIST, use

```
comp1 Description of 'comp1'
comp2 Description of 'comp2'
```

...

## Author(s)

 ${\bf Grant~Izmirlian < izmirlian@nih.gov}$ 

## See Also

```
objects to See Also as help,
```

```
## none as yet
```

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CY2TOShaz Co	alender yed	ar rates to	Study Year	Rates
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#### Description

Given the cutpoints at which the hazard is to be constant, the values taken by the calender year rates and the calender time offset from the start of the trial at which randomization ended, this function converts to time on study rates, assuming uniform accrual.

#### Usage

```
CY2TOShaz(tcut, t.eor, m, verbose = FALSE)
```

#### Arguments

tcut Left hand endpoints of intervals on which time on study hazard is taken

to be constant

t.eor Time offsest from the beginning of the trial at which randomization ended

m Annual calender time rates

verbose do you want to see alot of debugging info-defaults to FALSE

#### Value

```
hazard = h, table = attr(obj., "tbl")
```

hazard time on study hazard values taken on intervals specified by the argument

tcut

table a table containg the observed and fitted values

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

## Examples

```
## none as yet
```

CRRtoRR

Cumulative-risk ratios to risk ratios

#### Description

Given a vector of cumulative-risk ratios, computes risk ratios

## Usage

```
CRRtoRR(CRR, DT, h = NULL)
```

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#### Arguments

CRR vector of cumulative risk ratios of length m

DT vector of time increments upon which the cumulative ratios represent.

For example if the hazard takes values  $h_1,h_2,\ldots,h_m$  on the intervals  $[t_1,t_2),[t_2,t_3),\ldots,[t_m,t_{m+1})$  then DT will be  $\mathbf{c}(t_2-t_1,t_3-t_2,\ldots,t_{m+1}-t_m)$ 

 $t_m$ 

h The hazard in the reference arm, of length m

#### Value

The vector of risk ratios at the m time points

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

## Examples

```
## none as yet
```

RCM2RR

 $Relative\ cumulative\ mortality\ to\ Relative\ Risk$ 

#### Description

Relative cumulative mortality to Relative Risk

#### Usage

```
RCM2RR(tlook, tcut.i, h.i, hOth, accru, rcm)
```

## Arguments

tlook

tcut.i

h.i

h0th

accru

rcm

#### Value

Describe the value returned If it is a LIST, use

comp1 Description of 'comp1'
comp2 Description of 'comp2'

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

RR2RCM31

#### See Also

objects to See Also as help,

## Examples

## none as yet

RR2RCM

Relative risk to Relative Cumulative Mortality

## Description

Relative risk to Relative Cumulative Mortality

## Usage

```
RR2RCM(tlook, tcut.i, tcut.ii, h, rr, hOth, accru)
```

## Arguments

tlook	Describe tlook here
tcut.i	Describe tcut.i here
tcut.ii	Describe tcut.ii here
h	Describe h here
rr	Describe rr here
hOth	Describe h0th here
accru	Describe accru here

## Value

Describe the value returned If it is a LIST, use

Description of 'comp1' comp1 Description of 'comp2' comp2

## Author(s)

 ${\bf Grant~Izmirlian < izmirlian@nih.gov}$ 

#### See Also

objects to See Also as help,

```
## none as yet
```

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lookup Lookup values for a piecewise constant function
--

## Description

Given the values and lefthand endpoints for intervals of constancy, lookup values of the function at arbitrary values of the independent variable.

## Usage

```
lookup(xgrid, ygrid, x, y0 = 0)
```

## Arguments

xgrid	Lefthand endpoints of intervals of constancy
ygrid	Values on these intervals, of same length as xgrid
X	Input vector of arbitrary independent variables.
уO	Value to be returned for values of $x$ that are smaller than $min(xgrid)$ .

#### Value

Describe the value returned If it is a LIST, use

comp1	Description of 'comp1'
comp2	Description of 'comp2'

## Author(s)

Grant Izmirlian <izmirlian@nih.gov>

## Examples

```
## none as yet
```

${\tt lung} \qquad \qquad {\tt \it Mayo} \ {\tt \it Clinic} \ {\tt \it Lung} \ {\tt \it Cancer} \ {\tt \it Data}$
---

## Description

Survival in patients with lung cancer at Mayo Clinic. Performance scores rate how well the patient can perform usual daily activities.

## Usage

data(lung)

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#### **Format**

inst: Institution code time: Survival time in days

status: censoring status 1=censored, 2=dead

age: Age in years sex: Male=1 Female=2

ph.ecog: ECOG performance score (0=good 5=dead)

ph.karno: Karnofsky performance score (bad=0-good=100) rated by physician

pat.karno: Karnofsky performance score rated by patient

meal.cal: Calories consumed at meals wt.loss: Weight loss in last six months

#### Source

Terry Therneau

DX

function to do ...

#### Description

A concise (1-5 lines) description of what the function does.

## Usage

DX(x)

#### Arguments

X

Describe x here

#### **Details**

If necessary, more details than the description above

#### Value

Describe the value returned If it is a LIST, use

comp1 Description of 'comp1'
comp2 Description of 'comp2'

...

## Author(s)

Grant Izmirlian <izmirlian@nih.gov

34 paste

#### References

put references to the literature/web site here

## Examples

```
## none as yet
```

paste

The paste operator

## Description

A binary operator shortcut for paste(x,y)

## Usage

```
x %,% y
```

## Arguments

```
x a character stringy a character string
```

## Value

```
paste(x, y, sep="")
```

## Author(s)

Grant Izmirlian  $\langle izmirlian@nih.gov \rangle$ 

```
library(PwrGSD)
"var" %,% (1:10)
```

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