

Enhanced Survival Design with S+SeqTrial 3g

Overview

Under the null hypothesis, statistical information in the proportional hazards model is directly proportional to the number of events. This applies to the logrank test as well, because it corresponds to the score test of a hazard ratio in the PH model.

Sample size calculations in proportional hazards models thus are based on finding N_J , the maximal number of events to be observed at time τ_J when the last of J potential analyses would be performed (so J is the maximum possible number of analyses). Logistically, however, the clinical trialist needs to know A , the number of subjects to accrue.

In order to calculate the number of subjects needed, some assumptions must be made about the underlying distributions of time to event and time to dropout. In S+SeqTrial 3g, we allow exponential, Weibull and user-specified distributions for time to event and time to dropout, as well as piecewise constant models for hazard rates and dropout.

We also allow different choices for the time t_A of patient accrual, the time $t_f = \tau_J - t_A$ of additional follow-up after accrual of the last subject, and the accrual distribution of entry times. In particular, the accrual distribution can be specified as a beta distribution for the entry times (scaled to the interval $(0, t_A)$), as the number of subjects accrued in each unit of time, or as a user specified function, returning the individual entry times. The scaled beta distribution conveniently allows many accrual scenarios e.g. like slow early accrual followed by faster accrual later in the trial.

In practice the clinical trialist will consider a variety of accrual scenarios along with time to event and time to dropout distribution assumptions. Under each scenario, S+SEQTRIAL calculates the probability that a randomly chosen patient on the study might have an observed event. S+SEQTRIAL provides a command line function seqPHNSubjects to estimate the number of subjects to accrue using this model. This function can also be accessed from the Time Distns. tab of the dialog for the proportional hazards model. Notationally, we consider that the i^{th} patient would have independent entry, event, and dropout times

$E_i =$ Entry time of individual i (from start of study)

$T_i =$ Time to event for individual i (from time of entry)

$D_i =$ Time to dropout for individual i (from time of entry)

distributed according to cumulative distribution functions F_E , F_T , and F_D , respectively.

These times are then used to define the potentially censored observations at time τ according to

$$\begin{aligned}\tau &= \text{Time of analysis (from start of study)} \\ Y_i &= \text{Observation time for individual } i \text{ (from time of entry)} \\ \delta_i &= \text{Indicator of event}\end{aligned}$$

$$\begin{aligned}Y_i &= \max(\min(T_i, D_i, \tau - E_i), 0) \\ \delta_i &= \mathbf{1}_{[Y_i=T_i]}\end{aligned}$$

We then choose the value of A to satisfy

$$\begin{aligned}N_J &= A \times P(\delta_i = 1 | \tau_J, F_E, F_T, F_D) \\ P(\delta_i = 1 | \tau_J, F_E, F_T, F_D) &= P([E_i < \tau_J] \& [T_i < D_i] \& [T_i < \tau_J - E_i])\end{aligned}$$

In the clinical trial setting, we consider a two sample problem in which we compare the distribution of time to some event across a “treatment” group (denoted group 1) and a “control” group (denoted group 0). We have measurements of times to the event that are potentially right censored. We presume non-informative censoring may arise from two mechanisms: “administrative censoring” that is due to continued survival without an event at the time of data analysis, and “dropout” in which a subject might be lost to follow-up for other reasons at some time prior to an event being observed.

We consider two time scales:

- “calendar time” which is measured from the time the study starts (which is denoted time 0)
- “study time” which is measured for each patient from the time that patient is accrued to the study.

In either case, time is measured in arbitrary units of the user’s choosing (e.g., days, weeks, months, years). All references to time are presumed to be in the same units.

Subject Accrual

The maximal number of subjects to be accrued to the study is denoted A . This quantity may be a fixed value specified by the user or computed from other quantities specified by the user. Alternatively, it may be a random variable with a distribution specified directly by the user or a distribution derived from other quantities that are user specified.

Subject accrual will start at calendar time 0 and continue for t_A time units. This quantity may be a fixed value specified by the user or computed from other quantities that are user specified. When specified directly by the user, t_A will most often be an integer. Alternatively, it may be a random variable with a distribution derived from other quantities that are user specified.

The entry time for the i^{th} subject is denoted E_i . The entry times for all individuals are assumed to be totally independent. Note the following relationship:

$$0 \leq E_1 \leq E_2 \leq E_3 \leq \dots \leq E_A \leq t_A$$

We let a_k denote the number of subjects accrued between calendar times $k - 1$ and k . In the most general case, a_k can be a random variable with distributions specified as described below. The user will also have the option of specifying the exact values of the a_k ’s. For ease of specification, the user may specify fewer than t_A values, in which case the last specified value is carried forward to all future time intervals. Should a user specify more values than necessary, the extra values will be ignored.

We have the following relationships:

$$a_k = \sum_{i=1}^A \mathbf{1}_{[k-1 < E_i \leq k]}$$

$$A = \sum_{k=1}^{t_A} a_k$$

A user will specify two of t_A , A , and $(a_1, a_2, \dots, a_{t_A})$, and the third will be computed.

When computing a point prediction of the sample size to be accrued, the distribution of entry times can be characterized in several ways:

- Specification of a parametric conditional distribution for (E_i/t_A) based on a scaled beta distribution having shape parameters $\alpha > 0, \beta > 0$. In this conditional distribution,

$$F_E(ut_A) = P(E_i/t_A \leq u | \alpha, \beta) = \frac{1}{B(\alpha, \beta)} \int_0^u v^{\alpha-1} (1-v)^{\beta-1} dv$$

for $0 < u < 1$. Note that choosing $\alpha = 1, \beta = 1$ corresponds to uniform accrual between calendar times 0 and t_A .

- Specification of the exact number of subjects (a_1, a_2, \dots) to be accrued in each unit of time. For ease of specification, the user may specify fewer than t_A values, in which case the last specified value is carried forward to all future time intervals. Should a user specify more values than necessary, the extra values will be ignored. The individual entry times will then have uniform conditional distributions $(E_i | k-1 < E_i \leq k, a_k) \sim \mathcal{U}(k-1, k)$.
- Specification of a function `entry()` that will generate the individual entry times. The function should take arguments A, t_A , and additional parameters. The values of the individual entry times are given by $E_i = \text{entry}(A, t_A, \text{parameters})[i]$.

When computing an interval prediction of the sample size A to be accrued or the time t_A of subject accrual, there are also several methods for specifying the variability in accrual patterns:

- Specification of the mean rate of subject accrual ω and a parametric conditional distribution for (E_i/t_A) based on a beta distribution having parameters shape parameters $\alpha > 0, \beta > 0$. In this setting, a Poisson process $A(t) \sim \text{Poisson}(t\omega)$ can be used to allow variability in A or t_A , but not both. The individual entry times will then be computed using the beta distribution, as above.
- Specification of the mean number of subjects $(\omega_1, \omega_2, \dots)$ to be accrued in each unit of time. For ease of specification, the user may specify fewer than t_A values, in which case the last specified value is carried forward to all future time intervals. Should a user specify more values than necessary, the extra values will be ignored. The number of subjects accrued in each interval will then be randomly sampled according to a Poisson distribution $a_k \sim \text{Poisson}(\omega_k)$, and the individual entry times will then have uniform conditional distribution $(E_i | k-1 < E_i \leq k, a_k) \sim \mathcal{U}(k-1, k)$. When this approach is used either A or t_A will be specified, but not both.
- Specification of a function `entry()` that will generate the individual entry times. The function will take arguments A and t_A , and the value of the individual entry times will be determined by $E_i = \text{entry}(A, t_A)[i]$.

Treatment Assignment

Randomization scheme will be specified by an argument `ratio` signifying that size of treatment groups will be in the ratio `ratio[1]` on the treatment arm to `ratio[2]` on the control arm. (The default value for both `ratio[1]` and `ratio[2]` is 1).

Time to Event Distribution

The event time for the i^{th} subject is denoted T_i . The event times for all individuals are assumed to be totally independent.

The distribution for event times in the control group can be characterized several ways:

- Specification of a parametric Weibull distribution: $T \sim Weibull(p, \lambda_0)$. In this parametric distribution

$$F_T(t) = P(T \leq t) = 1 - \exp\left(-\left(\frac{t}{\lambda_0}\right)^p\right)$$
$$F_T^{-1}(u) = \lambda_0 (-\log(1 - u))^{1/p}$$

The user may specify either the values of the scale parameter `eventScale` = λ_0 and shape parameter `eventShape` = p (default $p = 1$) or up to two quantiles of the distribution of time to event using arguments `eventQuantiles` and `eventProbs`. In the latter specification, if only one quantile is supplied, an exponential distribution is presumed ($p = 1$). The default value for `eventProbs` is 0.5 (so the median).

- Specification of a constant hazard ($\lambda_1, \lambda_2, \dots$) for events in each unit of time. For ease of specification, the user may specify fewer than $t_A + t_f$ values, in which case the last specified value is carried forward to all future time intervals. Should a user specify more values than necessary, the extra values will be ignored. The cumulative distribution function for the event times is then

$$F_T(t) = P(T \leq t) = 1 - \exp(-\Lambda(t))$$
$$\Lambda(t) = \sum_{k=1}^{\lfloor t \rfloor} \lambda_k + \lambda_{\lfloor t \rfloor} (t - \lfloor t \rfloor)$$

- Specification of a function `event()` that will generate the individual event times. The function will take arguments A , and the value of the individual entry times will be determined by $T_i = \text{event}(A)[i]$. (Note: A very common situation will be one in which the user wants to resample from a Kaplan-Meier curve derived from an earlier study.)

Time to Dropout Distribution

The dropout time for the i^{th} subject is denoted D_i . The event times for all individuals are assumed to be totally independent.

The distribution for dropout times can be characterized several ways:

- Specification of a parametric Weibull distribution: $D \sim Weibull(p, \gamma_0)$. In this parametric distribution

$$F_D(t) = P(D \leq t) = 1 - \exp\left(-\left(\frac{t}{\gamma_0}\right)^p\right)$$
$$F_D^{-1}(u) = \gamma_0 (-\log(1 - u))^{1/p}$$

The user may specify either the values of the scale parameter `dropoutScale` = γ_0 and shape parameter `dropoutScale` = p (default $p = 1$) or up to two quantiles of the distribution of time to dropout using arguments `dropoutQuantiles` and `dropoutProbs`. In the latter specification, if only one quantile is supplied, an exponential distribution is presumed ($p = 1$). The default value for `dropoutProbs` is 0.5 (so the median).

- Specification of a constant hazard ($\gamma_1, \gamma_2, \dots$) for dropouts in each unit of time. For ease of specification, the user may specify fewer than $t_A + t_f$ values, in which case the last specified value is carried forward to all future time intervals. Should a user specify more values than necessary, the extra values will be ignored. The cumulative distribution function for the dropout times is then

$$F_D(t) = P(D \leq t) = 1 - \exp(-\Lambda(t))$$

$$\Lambda(t) = \sum_{k=1}^{\lfloor t \rfloor} \gamma_k + \gamma_{\lceil t \rceil} (t - \lfloor t \rfloor)$$

- Specification of a function `dropout()` that will generate the individual dropout times. The function will take arguments A , and the value of the individual dropout times will be determined by $D_i = \text{dropout}(A)[i]$. (Note: A very common situation will be one in which the user wants to resample from a Kaplan-Meier curve derived from an earlier study.)

Some examples using seqPHNSubjects

Examples numbers such as **A8** in this section refer to the `Examples06.ssc` file in the `seqtrial/scripts/` folder in `S+SeqTrial`. See that file for output abridged here, and for additional examples. **A** examples focus on accrual, **E** on event times, **D** on dropout times, **T** on theta parameters accrual summary printouts, **O** on operating characteristics, and **P** on power.

I'm not sure about P.

Example A8: User provides $A; (a_1, a_2, \dots)$

You plan for $A = 700$ subjects, with accrual expected to ramp linearly up to 30 subjects per month (time unit) over the first 6 months and remain at 30 subjects per month thereafter: $a_1 = 5, a_2 = 10, \dots, a_6 = 30, a_7 = \dots = 30$.

- t_A is determined as the number of time units required to achieve the desired sample size A with the given rates per time unit, i.e. such that

$$\sum_{i=1}^{t_A} a_i = A$$

- Cumulative distribution function for entry times:

$$F_E(u) = P(E_i \leq u) = \frac{\sum_{i=1}^{\lfloor u \rfloor} a_i + a_{\lceil u \rceil} (u - \lfloor u \rfloor)}{\sum_{i=1}^{t_A} a_i}$$

- We will compute the accrual time and study time (and the analysis times, $\tau_1, \tau_2, \dots, \tau_J$) necessary to see the desired number of events; assuming exponential time to event distribution and no dropout.

Summary:

- **Accrual:** Specify `accrualSize`, `accrualRate`
`accrualTime`, `studyTime` are computed
- **Event:** `eventQuantiles` = 12 (median for exponential distribution)
- **Dropout:** none (default)

Command-line Code

```
dsnA8 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,  
                      P=c(1, 0.5), accrualSize=700,  
                      accrualRate=c(5, 10, 15, 20, 25, 30),  
                      eventQuantiles=12, seed=0)  
  
dsnA8
```

Output

```
PROBABILITY MODEL and HYPOTHESES:  
Two arm study of censored time to event response variable  
Theta is hazard ratio (Treatment : Comparison)  
One-sided hypothesis test of a lesser alternative:  
    Null hypothesis : Theta >= 1      (size = 0.025)  
    Alternative hypothesis : Theta <= 0.7 (power = 0.975)
```

```
STOPPING BOUNDARIES: Sample Mean scale  
                   a      d  
Time 1 (NEv= 119.04) 0.4499 1.0872  
Time 2 (NEv= 238.08) 0.6707 0.9557  
Time 3 (NEv= 357.11) 0.7662 0.9026  
Time 4 (NEv= 476.15) 0.8190 0.8724  
Time 5 (NEv= 595.19) 0.8523 0.8523
```

```
ACCRUAL INFORMATION:  
Accrual distribution:  
    Poisson process  
    Number of subjects:      700  
    Accrual time      :      25.83333  
    Accrual rate      :      5 10 15 20 25 30  
Event distribution:  
    Exponential (hazard=0.05776227; 50th %ile=12)  
Dropout distribution:  
    No Dropout
```

```
Accrual summary table:  
      theta Scenario NAccrual accrualRate accrualTime studyTime  
alternative 0.7      1      700          NA      25.83      54.89  
      null 1.0      1      700          NA      25.83      48.51
```

Timing of analyses:

```
Theta = 0.7      Scenario 1
```

	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	16.66	23.56	29.82	38.69	54.89
N Accrued	424.23	631.49	700.00	700.00	700.00
N Events	119.04	238.08	357.11	476.15	595.19

Theta = 1	Scenario 1	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time		15.77	22.21	27.88	35.28	48.51
N Accrued		398.14	591.28	700.00	700.00	700.00
N Events		119.04	238.08	357.11	476.15	595.19

Example A5: User provides $t_A; (a_1, a_2, \dots)$

You plan for $t_A = 24$ months (time units) for accrual, with accrual expected to ramp linearly up to 30 subjects per month (time unit) over the first 6 months and remain at 30 subjects per month thereafter: $a_1 = 5, a_2 = 10, \dots, a_6 = 30, a_7 = \dots = 30$.

- A is determined as $A = \sum_{i=1}^{t_A} a_i$.
- Cumulative distribution function for entry times:

$$F_E(u) = P(E_i \leq u) = \frac{\sum_{i=1}^{\lfloor u \rfloor} a_i + a_{\lceil u \rceil} (u - \lfloor u \rfloor)}{\sum_{i=1}^{t_A} a_i}$$

- We will compute the study time t_f (and the analysis times $\tau_1, \tau_2, \dots, \tau_J$) necessary to see the desired number of events; assuming exponential time to event distribution and no dropout.

Summary:

- **Accrual:** Specify `accrualTime`, `accrualRate`
`studyTime` is computed
- **Event:** `eventQuantiles` = 12 (median for exponential distribution)
- **Dropout:** none (default)

Command-line Code

```
dsnA5 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24,
  accrualRate=c(5, 10, 15, 20, 25, 30),
  eventQuantiles=12, seed=0)
dsnA5
```

Output

```
PROBABILITY MODEL and HYPOTHESES:
Two arm study of censored time to event response variable
Theta is hazard ratio (Treatment : Comparison)
One-sided hypothesis test of a lesser alternative:
Null hypothesis : Theta >= 1      (size = 0.025)
Alternative hypothesis : Theta <= 0.7  (power = 0.975)
```

STOPPING BOUNDARIES: Sample Mean scale

	a	d
Time 1 (NEv= 119.04)	0.4499	1.0872
Time 2 (NEv= 238.08)	0.6707	0.9557
Time 3 (NEv= 357.11)	0.7662	0.9026
Time 4 (NEv= 476.15)	0.8190	0.8724
Time 5 (NEv= 595.19)	0.8523	0.8523

ACCRUAL INFORMATION:

Accrual distribution:

Poisson process

Number of subjects: 645

Accrual time : 24

Accrual rate : 5 10 15 20 25 30

Event distribution:

Exponential (hazard=0.05776227; 50th %ile=12)

Dropout distribution:

No Dropout

Accrual summary table:

	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	645	NA	24	68.24
null	1.0	1	645	NA	24	58.80

Timing of analyses:

Theta = 0.7	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	16.67	23.57	30.67	41.89	68.24
N Accrued	424.31	632.00	645.00	645.00	645.00
N Events	119.04	238.08	357.11	476.15	595.19

Theta = 1	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	15.78	22.22	28.38	37.62	58.8
N Accrued	398.84	591.78	645.00	645.00	645.0
N Events	119.04	238.08	357.11	476.15	595.2

Example A2: User provides $A; t_A; \alpha, \beta$

You plan for $A = 700$ subjects, with $t_A = 24$ months (time units) for accrual. You expect a distribution of accrual times that is approximately beta with $\alpha = 10, \beta = 1$ (so accrual rates increase as the study progresses).

- Cumulative distribution function for entry times:

$$F_E(ut_A) = P(E_i/t_A \leq u | \alpha, \beta) = \frac{1}{B(\alpha, \beta)} \int_0^u \left(\frac{v}{t_A}\right)^{\alpha-1} \left(1 - \frac{v}{t_A}\right)^{\beta-1}$$

- We will compute the study time t_f (and the analysis times $\tau_1, \tau_2, \dots, \tau_J$) necessary to see the desired number of events; assuming exponential time to event distribution and no dropout.

Summary:

- **Accrual:** Specify `accrualSize`, `accrualTime`, `aShapeAccr`
`aShapeAccr=10` is faster accrual late,
`studyTime` is computed.
- **Event:** `eventQuantiles = 12` (median for exponential distribution)
- **Dropout:** none (default)

Command-line Code

```
dsnA2 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,  
P=c(1, 0.5), accrualSize=700, accrualTime=24, aShapeAccr=10,  
eventQuantiles=12, seed=0)  
dsnA2
```

Output

PROBABILITY MODEL and HYPOTHESES:

```
Two arm study of censored time to event response variable  
Theta is hazard ratio (Treatment : Comparison)  
One-sided hypothesis test of a lesser alternative:  
    Null hypothesis : Theta >= 1      (size = 0.025)  
    Alternative hypothesis : Theta <= 0.7 (power = 0.975)
```

STOPPING BOUNDARIES: Sample Mean scale

```
          a      d  
Time 1 (NEv= 119.04) 0.4499 1.0872  
Time 2 (NEv= 238.08) 0.6707 0.9557  
Time 3 (NEv= 357.11) 0.7662 0.9026  
Time 4 (NEv= 476.15) 0.8190 0.8724  
Time 5 (NEv= 595.19) 0.8523 0.8523
```

ACCRUAL INFORMATION:

```
Accrual distribution:  
    Beta distribution (shape1=10, shape2=1)  
    Number of subjects:      700  
    Accrual time      :      24  
    Accrual rate      :      Unspecified  
Event distribution:  
    Exponential (hazard=0.05776227; 50th %ile=12)  
Dropout distribution:  
    No Dropout
```

Accrual summary table:

	theta	Scenario	N	accrualRate	accrualTime	studyTime
alternative	0.7	1	700	NA	24	61.49
null	1.0	1	700	NA	24	54.78

Timing of analyses:

```
Theta = 0.7    Scenario 1
```

	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	25.71	30.35	36.54	45.44	61.49
N Accrued	700.00	700.00	700.00	700.00	700.00
N Events	119.04	238.08	357.11	476.15	595.19

Theta = 1	Scenario 1	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time		25.16	29.15	34.26	41.65	54.78
N Accrued		700.00	700.00	700.00	700.00	700.00
N Events		119.04	238.08	357.11	476.15	595.19

Example A7: User provides $A; t_A; \text{entry}()$

You plan for $A = 700$ subjects, with $t_A = 24$ months (time units) for accrual. You have a function `entry()` that returns times according to some distribution that you expect for accrual times.

- Cumulative distribution function for entry times will be approximated using a large sample of accrual times from the user-supplied function `entry()`.
- We will compute the study time t_f (and the analysis times $\tau_1, \tau_2, \dots, \tau_J$) necessary to see the desired number of events; assuming exponential time to event distribution and no dropout.

Summary:

- **Accrual:** Specify `accrualTime`, `accrualSize`, `entry`, `entryArgs`
`studyTime` is computed.
- **Event:** `eventQuantiles = 12` (median for exponential distribution)
- **Dropout:** `none` (default)

Command-line Code

```
entryA7 <- function(n, time, entryArgs) {
  distn <- entryArgs$distn
  distn <- distn - min(distn)
  distn <- distn / max(distn)
  sample (distn, n, replace=T) * time
}
set.seed(1)
entryArgsA7 <- list(distn=rbeta(1000, 0.5, 0.5))
dsnA7 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, accrualSize=700, entry=entryA7,
  entryArgs=entryArgsA7, eventQuantiles=12, seed=0)
dsnA7
```

Output

PROBABILITY MODEL and HYPOTHESES:

```
Two arm study of censored time to event response variable
Theta is hazard ratio (Treatment : Comparison)
One-sided hypothesis test of a lesser alternative:
  Null hypothesis : Theta >= 1      (size = 0.025)
  Alternative hypothesis : Theta <= 0.7  (power = 0.975)
```

STOPPING BOUNDARIES: Sample Mean scale

	a	d
Time 1 (NEv= 119.04)	0.4499	1.0872
Time 2 (NEv= 238.08)	0.6707	0.9557
Time 3 (NEv= 357.11)	0.7662	0.9026
Time 4 (NEv= 476.15)	0.8190	0.8724
Time 5 (NEv= 595.19)	0.8523	0.8523

ACCRUAL INFORMATION:

Accrual distribution:

User-specified function: entryA7

Number of subjects: 700

Accrual time : 24

Accrual rate : Unspecified

Event distribution:

Exponential (hazard=0.05776227; 50th %ile=12)

Dropout distribution:

No Dropout

Accrual summary table:

	theta	Scenario	N	accrualRate	accrualTime	studyTime
alternative	0.7	1	700	NA	24	53.77
null	1.0	1	700	NA	24	47.37

Timing of analyses:

Theta = 0.7	Scenario	1			
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	13.6	22.21	28.73	37.6	53.77
N Accrued	371.2	562.23	700.00	700.0	700.00
N Events	119.0	238.08	357.11	476.2	595.19

Theta = 1	Scenario	1			
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	12.47	20.64	26.8	34.16	47.37
N Accrued	350.01	519.59	700.0	700.00	700.00
N Events	119.04	238.08	357.1	476.15	595.19

Example: User provides $t_A; \alpha, \beta; \tau_J$

This is similar to example **A2**, except that instead of specifying the total number of accrued we specify the total study time.

You plan for $t_A = 24$ months (time units) for accrual and $\tau_J = 36$ months (time units) total for the study, including the accrual period. You expect a distribution of accrual times that is approximately beta with $\alpha = 10, \beta = 1$ (as for example **A2**)

- We will compute the sample size A so that by time $\tau_J = t_A + t_f$ we expect to have seen the desired number of events.
- We will compute the sample size A so that by time $\tau_J = t_A + t_f$ we expect to have seen the desired number of events. In comparison to example **A2**, where the study time was 61.49 under the alternative

hypothesis, here the study time is much shorter, only 36. This allows less time for followup, and as a result the total accrual must be much larger, 1197 under the alternative hypothesis instead of 700.

Summary:

- **Study time:** Specify `studyTime`
- **Accrual:** Specify `accrualTime`
uniform accrual (default)
`accrualSize` is computed
- **Event:** `eventQuantiles = 12` (median for exponential distribution)
- **Dropout:** none (default)

Command-line Code

```
dsnA <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,  
  P=c(1, 0.5), accrualTime=24, aShapeAccr=10, studyTime = 36,  
  eventQuantiles=12, seed=0)  
dsnA
```

Output

```
PROBABILITY MODEL and HYPOTHESES:  
  Two arm study of censored time to event response variable  
  Theta is hazard ratio (Treatment : Comparison)  
  One-sided hypothesis test of a lesser alternative:  
    Null hypothesis : Theta >= 1      (size = 0.025)  
    Alternative hypothesis : Theta <= 0.7  (power = 0.975)  
  
STOPPING BOUNDARIES: Sample Mean scale  
      a      d  
Time 1 (NEv= 119.04) 0.4499 1.0872  
Time 2 (NEv= 238.08) 0.6707 0.9557  
Time 3 (NEv= 357.11) 0.7662 0.9026  
Time 4 (NEv= 476.15) 0.8190 0.8724  
Time 5 (NEv= 595.19) 0.8523 0.8523  
  
ACCRUAL INFORMATION:  
  Accrual distribution:  
    Beta distribution (shape1=10, shape2=1)  
    Number of subjects:      Unspecified  
    Accrual time      :      24  
    Accrual rate      :      Unspecified  
  Event distribution:  
    Exponential (hazard=0.05776227; 50th %ile=12)  
  Dropout distribution:  
    No Dropout
```

Accrual summary table:

	theta	Scenario	N	accrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	1197	NA	24	36	

null	1.0	1	1071	NA	24	36
------	-----	---	------	----	----	----

Timing of analyses:

Theta = 0.7	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	24.02	26.42	29.11	32.22	36.0
N Accrued	1197.00	1197.00	1197.00	1197.00	1197.0
N Events	119.04	238.08	357.11	476.15	595.2

Theta = 1	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	23.98	26.3	28.98	32.07	36.0
N Accrued	1062.62	1071.0	1071.00	1071.00	1071.0
N Events	119.04	238.1	357.11	476.15	595.2

Example A14: User provides t_A

You plan for $t_A = 24$ months (time units) accrual, with uniform accrual.

In contrast to the examples above, this example is incompletely specified; various combinations of accrual size and study time will provide the required number of events.

Also, in this example we use graphical procedures to compare accrual plans.

- We will consider a range of accrual sizes A (or equivalently, accrual rates a).
- We will compute the study time t_f and (analysis times $\tau_1, \tau_2, \dots, \tau_J$) necessary to see the desired number of events; assuming exponential time to event distribution and no dropout.

Summary:

- **Accrual:** Specify `accrualTime`
uniform accrual (default)
range of `accrualSize`, `studyTime` is computed
- **Event:** `eventQuantiles = 12` (median for exponential distribution)
- **Dropout:** none (default)

Command-line Code

```
dsnA14 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7, P=c(1,0.5),
  accrualTime=24, eventQuantiles=12, seed=0)
dsnA14

seqPlotAccrualFupTime(dsnA14)

dsnA14b <- update(dsnA14, accrualTime = 28)
seqPlotAccrualFupTime(dsnA14, dsnA14b)

## Test combination of seqDesignExtd and seqDesign
dsnA14d <- seqDesign("hazard", nbr=5, test.type="less", alt=0.7,P=c(1,0.5))
seqPlotAccrualFupTime(dsnA14, dsnA14d)
# Excludes dsnA14d and gives warning (as expected)

seqPlotPHNSubjects(dsnA14)
```

Output (abridged)

PROBABILITY MODEL and HYPOTHESES:

Two arm study of censored time to event response variable

Theta is hazard ratio (Treatment : Comparison)

One-sided hypothesis test of a lesser alternative:

Null hypothesis : $\Theta \geq 1$ (size = 0.025)

Alternative hypothesis : $\Theta \leq 0.7$ (power = 0.975)

STOPPING BOUNDARIES: Sample Mean scale

	a	d
Time 1 (NEv= 119.04)	0.4499	1.0872
Time 2 (NEv= 238.08)	0.6707	0.9557
Time 3 (NEv= 357.11)	0.7662	0.9026
Time 4 (NEv= 476.15)	0.8190	0.8724
Time 5 (NEv= 595.19)	0.8523	0.8523

ACCRUAL INFORMATION:

Accrual distribution:

Uniform distribution

Number of subjects: Unspecified

Accrual time : 24

Accrual rate : Unspecified

Event distribution:

Exponential (hazard=0.05776227; 50th %ile=12)

Dropout distribution:

No Dropout

Accrual summary table:

	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	596.0	24.83	24	158.93
alternative	0.7	2	690.8	28.78	24	54.52
alternative	0.7	3	785.6	32.73	24	42.52
alternative	0.7	4	880.3	36.68	24	36.46
alternative	0.7	5	975.1	40.63	24	32.56
alternative	0.7	6	1069.9	44.58	24	29.82
alternative	0.7	7	1164.7	48.53	24	27.83
alternative	0.7	8	1259.4	52.48	24	26.24
alternative	0.7	9	1354.2	56.43	24	25.02
alternative	0.7	10	1449.0	60.38	24	23.99
null	1.0	1	596.0	24.83	24	129.39
null	1.0	2	673.2	28.05	24	50.57
...						
null	1.0	10	1291.0	53.79	24	23.99

Timing of analyses:

Theta = 0.7	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	15.76	23.59	31.97	46.53	158.9
N Accrued	392.82	585.59	596.00	596.00	596.0
N Events	119.04	238.08	357.11	476.15	595.2

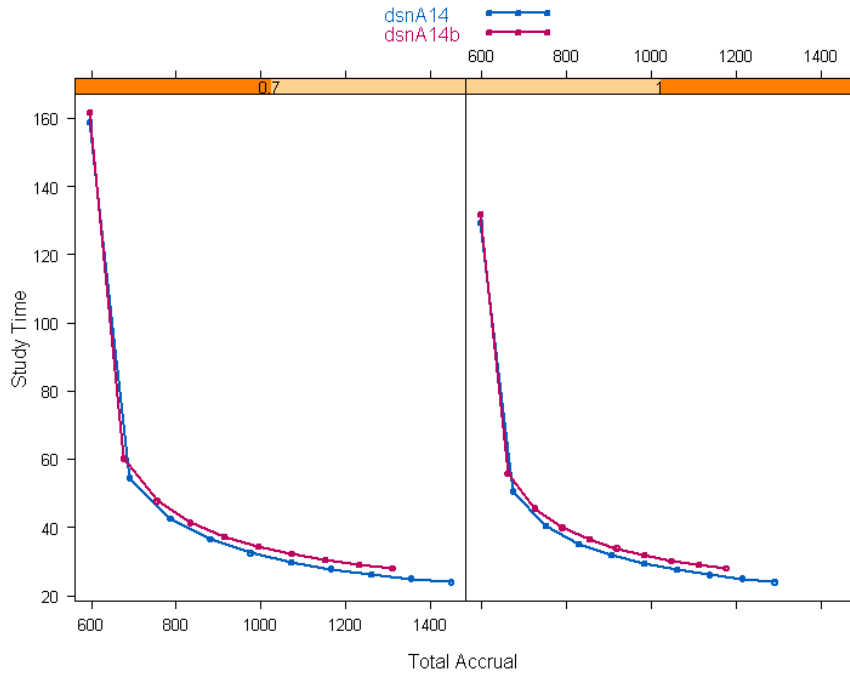
```

Theta = 0.7 Scenario 2
Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time 14.55 21.59 28.09 37.28 54.52
N Accrued 420.12 622.37 690.78 690.78 690.78
N Events 119.04 238.08 357.11 476.15 595.19
...
Theta = 0.7 Scenario 10
Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time 9.713 14.16 17.71 20.97 23.99
N Accrued 587.309 857.79 1073.06 1268.05 1448.29
N Events 119.038 238.08 357.11 476.15 595.19

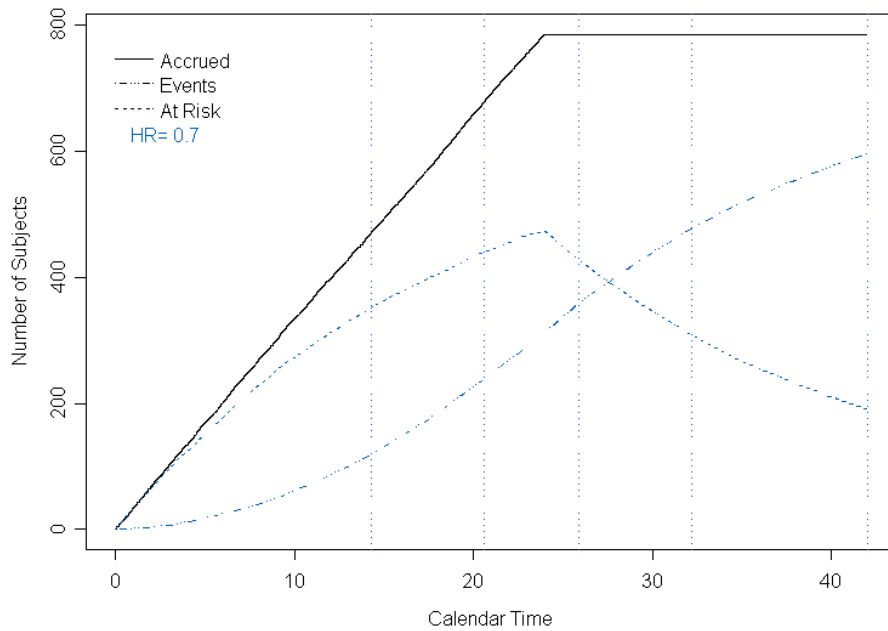
Theta = 1 Scenario 1
Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time 14.73 22.02 29.1 41.08 129.4
N Accrued 366.96 547.00 596.0 596.00 596.0
N Events 119.04 238.08 357.1 476.15 595.2
...
Theta = 1 Scenario 10
Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time 9.524 14.05 17.75 21.02 23.99
N Accrued 512.979 757.91 957.21 1132.08 1290.32
N Events 119.038 238.08 357.11 476.15 595.19

```

use selected figures here



Scenario 3 Designed for Theta 0.7



Example A15: User provides (a_1, a_2, \dots)

You plan for accrual expected to ramp linearly up to 30 subjects per month (time unit) over the first 6 months and remain at 30 subjects per month thereafter: $a_1 = 5, a_2 = 10, \dots, a_6 = 30, a_7 = \dots = 30$.

This example is incompletely specified; various combinations of accrual size (or accrual time) and study time will provide the required number of events.

- We will consider a range of accrual sizes A (or equivalently, total accrual times t_A).
- We will compute the study time t_f and (analysis times $\tau_1, \tau_2, \dots, \tau_J$) necessary to see the desired number of events; assuming exponential time to event distribution and no dropout.

Summary:

- **Accrual:** Specify `accrualRate`
range of `accrualSize`, `studyTime` is computed
- **Event:** `eventQuantiles = 12` (median for exponential distribution)
- **Dropout:** none (default)

Command-line Code

```
dsnA15 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7, P=c(1,0.5),
                        accrualRate=c(5,10,15,20,25,30), eventQuantiles=12, seed=0)
dsnA15
```


Output (abridged)

PROBABILITY MODEL and HYPOTHESES:

Two arm study of censored time to event response variable

Theta is hazard ratio (Treatment : Comparison)

One-sided hypothesis test of a lesser alternative:

Null hypothesis : $\Theta \geq 1$ (size = 0.025)

Alternative hypothesis : $\Theta \leq 0.7$ (power = 0.975)

STOPPING BOUNDARIES: Sample Mean scale

	a	d
Time 1 (NEv= 119.04)	0.4499	1.0872
Time 2 (NEv= 238.08)	0.6707	0.9557
Time 3 (NEv= 357.11)	0.7662	0.9026
Time 4 (NEv= 476.15)	0.8190	0.8724
Time 5 (NEv= 595.19)	0.8523	0.8523

ACCRUAL INFORMATION:

Accrual distribution:

Poisson process

Number of subjects: Unspecified

Accrual time : Unspecified

Accrual rate : 5 10 15 20 25 30

Event distribution:

Exponential (hazard=0.05776227; 50th %ile=12)

Dropout distribution:

No Dropout

Accrual summary table:

	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	596.0	NA	22.37	167.91
alternative	0.7	2	653.0	NA	24.27	74.63
...						
alternative	0.7	10	1108.6	NA	39.45	39.47
null	1.0	1	596.0	NA	22.37	138.66
...						
null	1.0	10	1041.9	NA	37.23	37.34

Timing of analyses:

Theta = 0.7	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	22.7	33.18	42.56	56.99	167.9
N Accrued	326.0	494.98	596.00	596.00	596.0
N Events	119.0	238.08	357.11	476.15	595.2

Theta = 0.7	Scenario 2				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	21.68	31.47	39.94	50.69	74.63
N Accrued	339.12	512.53	652.96	652.96	652.96
N Events	119.04	238.08	357.11	476.15	595.19

...

```

Theta = 0.7   Scenario 10
              Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time    16.62    23.61    29.35    34.56    39.47
  N Accrued     422.29    633.41    806.76    961.94   1108.62
  N Events      119.04    238.08    357.11    476.15    595.19

Theta = 1     Scenario 1
              Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time    20.61    30.12    38.51    50.44   138.7
  N Accrued     310.54    473.90    596.00    596.00    596.0
  N Events      119.04    238.08    357.11    476.15    595.2
...
Theta = 1     Scenario 10
              Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time    15.68    22.12    27.65    32.62    37.34
  N Accrued     395.42    588.06    753.71    904.07   1041.93
  N Events      119.04    238.08    357.11    476.15    595.19

```

Examples E1–E5: Specifying Event Distributions

These examples demonstrate different ways to specify event distributions—two ways to specify Weibull distributions, an example with piecewise exponential hazard distributions, an example with a custom event function, and an example using pilot data.

The examples have the same accrual specification — accrual time of 24 months and study time of 60 months, and uniform accrual rate (except accrual and study times are modified for the pilot data example).

- Uniform accrual over 24 months, study time 60 months
- **E1:** specify Weibull distribution using quantiles
- **E2:** specify Weibull distribution using shape parameters
- **E3:** specify piecewise exponential distribution
- **E4:** define a function for event times.
- **E5:** use pilot data (using `eventPilot` as an event function).

Summary:

- **Accrual:** Specify `accrualTime`, `studyTime`
entry distribution is Uniform, sample size is computed.
- **Dropout:** none (default)
- **Event E1:** `eventQuantiles = 12, 24`, `eventProbs = 0.5, 0.75` (Weibull)
- **Event E2:** `eventScale = 0.1` `eventShape = 2` (Weibull)
- **Event E3:** `eventHazard = 0.01, 0.02, 0.03, 0.04, 0.10, 0.5, 0.05` (piecewise exponential)
- **Event E4:** user-defined function
- **Event E5:** pilot function

Command-line Code

```
dsnE1 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, seed=0,
  eventQuantiles=c(12, 24), eventProbs=c(0.5, 0.75))

dsnE2 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, seed=0,
  eventScale=0.1, eventShape=2)

dsnE3 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, seed=0,
  eventHazard=c(1:4, 10, 50, 5)/100)
seqPlotPHNSubjects(dsnE3, theta=c(0.7,1))

eventE4 <- function(n, nTrt, theta, eventArgs, pcwsHR) {
  distn <- eventArgs$distn
  distn <- sort(distn)
  pdistn.ctrl <- 1-c(0, (1:length(distn)) / length(distn))
  pdistn.treat <- pdistn.ctrl ^ theta
  nctrl <- n - nTrt
  c(as.numeric(distn[cut(runif(nTrt), 1-pdistn.treat)]),
    as.numeric(distn[cut(runif(nctrl), 1-pdistn.ctrl)]))
}
set.seed(1)
eventArgsE4 <- list(distn=rlnorm(1000, 3, 1))
x <- (1:999)/1000
plot(qlnorm(1-x, 3, 1), x, type="l")
lines(qlnorm(1-x, 3, 1), x^0.7)
plot(qlnorm(1-x, 3, 1), x, type="l", xlim=c(0, 60))
lines(qlnorm(1-x, 3, 1), x^0.7)
z <- eventE4(2000, 1000, 0.7, eventArgsE4) # This should be run after
# defining eventArgsE4, to keep the same random numbers as for the
# results below.
plot(survfit(Surv(z, rep(1, 2000))~rep(1:0, each=1000)))
dsnE4 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, event=eventE4,
  eventArgs=eventArgsE4, seed=0)
dsnE4
seqPlotPHNSubjects(dsnE4, theta=c(0.7,1))

# See Examples06.ssc for commands used to create the pilot data set "Lifetest"
plot(survfit(Time ~ Treatment, data = Lifetest))
random <- eventPilot( 190, 100, Lifetest) # random data
points(sort(random$treatment), 100:1/100, col=2)
points(sort(random$control), 90:1/90, col=3)

dsnE5 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
P=c(1, 0.5), accrualTime=240, studyTime=300, event=eventPilot,
eventArgs=list(data = Lifetest), seed=0)
dsnE5
plot(dsnE5)
```

seqPlotPHNSubjects(dsnE5)

Output Output from example E4 is shown; the others are similar.

PROBABILITY MODEL and HYPOTHESES:

Two arm study of censored time to event response variable
Theta is hazard ratio (Treatment : Comparison)
One-sided hypothesis test of a lesser alternative:
Null hypothesis : $\Theta \geq 1$ (size = 0.025)
Alternative hypothesis : $\Theta \leq 0.7$ (power = 0.975)

STOPPING BOUNDARIES: Sample Mean scale

	a	d
Time 1 (NEv= 119.04)	0.4499	1.0872
Time 2 (NEv= 238.08)	0.6707	0.9557
Time 3 (NEv= 357.11)	0.7662	0.9026
Time 4 (NEv= 476.15)	0.8190	0.8724
Time 5 (NEv= 595.19)	0.8523	0.8523

ACCRUAL INFORMATION:

Accrual distribution:
Uniform distribution
Number of subjects: Unspecified
Accrual time : 24
Accrual rate : Unspecified
Event distribution:
User-specified function: eventE4
Dropout distribution:
No Dropout

Accrual summary table:

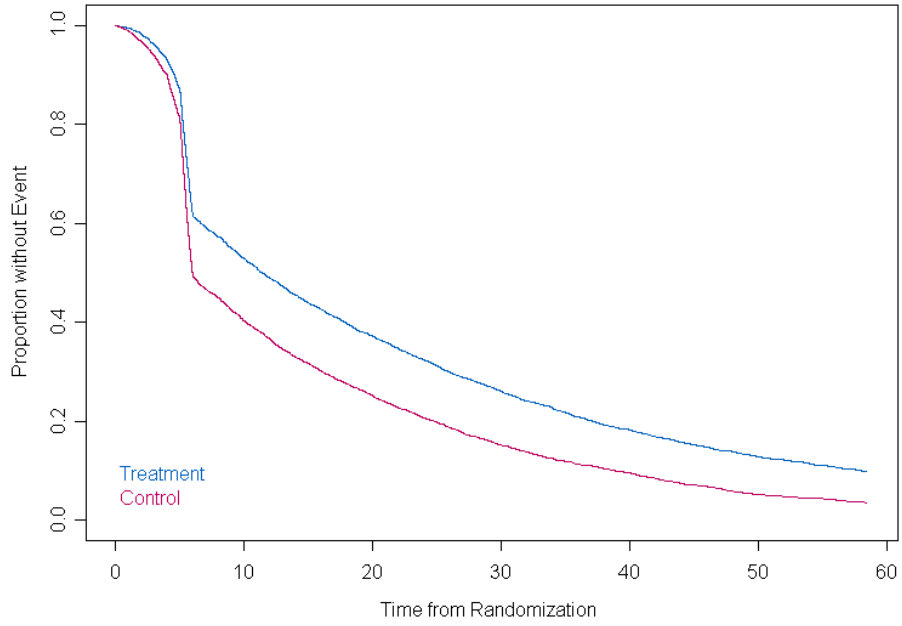
	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	796	33.17	24	60
null	1.0	1	734	30.58	24	60

Timing of analyses:

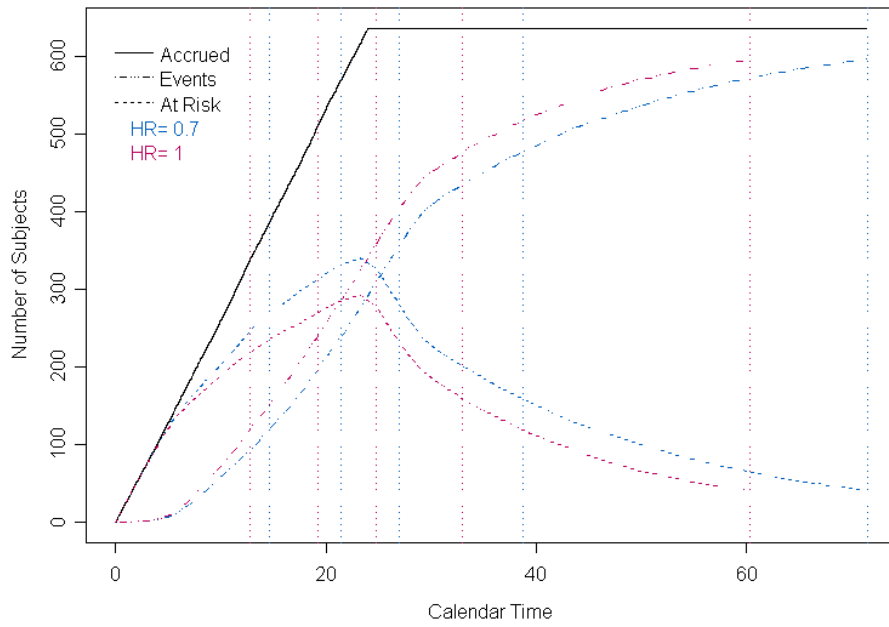
Theta = 0.7	Scenario	1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5	
Analysis Time	17.91	25.27	32.32	42.32	60.0	
N Accrued	595.65	796.00	796.00	796.00	796.0	
N Events	119.04	238.08	357.11	476.15	595.2	

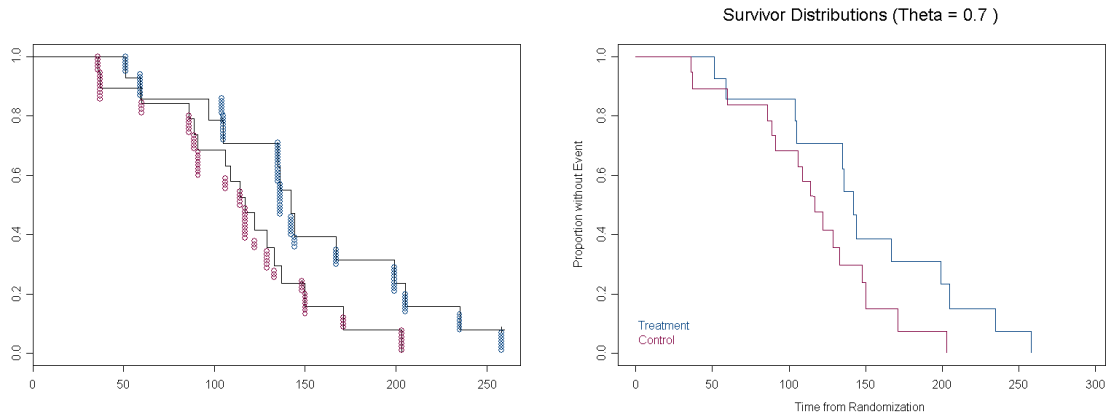
Theta = 1	Scenario	1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5	
Analysis Time	17.43	24.64	31.28	41.06	60.0	
N Accrued	534.69	734.00	734.00	734.00	734.0	
N Events	119.04	238.08	357.11	476.15	595.2	

Survivor Distributions (Theta = 0.7)

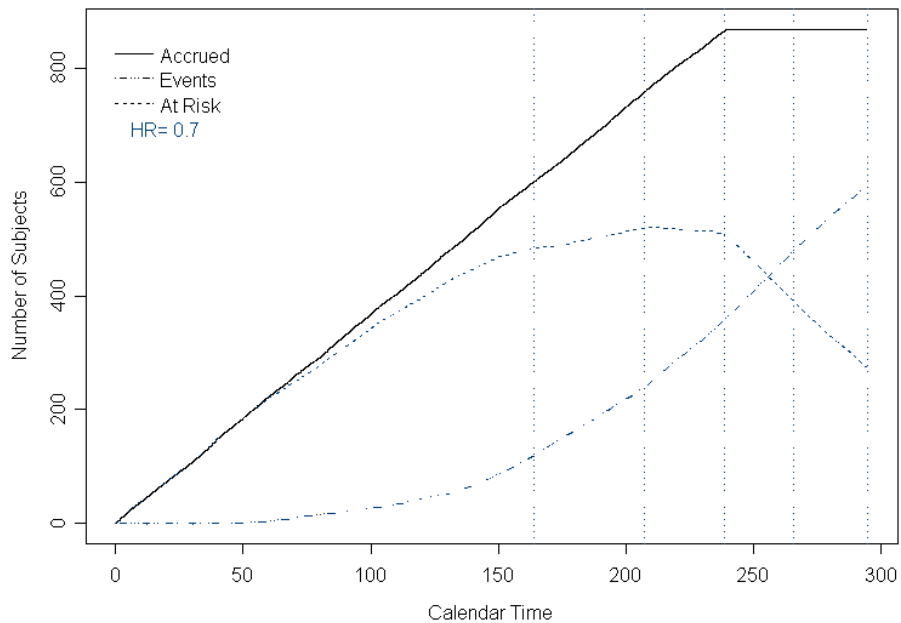


Scenario 1 Designed for Theta 1





Scenario 1 Designed for Theta 0.7



Examples D1–D6: Specifying Dropout Distributions

These examples demonstrate different ways to specify dropout distributions—exponential or Weibull distributions, piecewise exponential, or using a user-defined function.

The examples have the same accrual specification — accrual time of 24 months and study time of 60 months, and uniform accrual rate. And they share a common event time distribution, exponential with median of 12 months.

- Uniform accrual over 24 months, study time 60 months
- **D1:** specify exponential distribution using median
- **D2:** specify Weibull distribution using quantiles
- **D3:** specify exponential distribution using rate

- **D4:** specify piecewise exponential distribution
- **D5:** specify Weibull distribution using parameters
- **D6:** define a function for event times.

Summary:

- **Accrual:** Specify `accrualTime`, `studyTime`
entry distribution is Uniform, sample size is computed.
- **Event:** exponential with median 12
- **Dropout D1:** `dropoutQuantiles = 120` (exponential)
- **Dropout D2:** `dropoutQuantiles = 120, 240, dropoutProbs = 0.5 0.75` (Weibull)
- **Dropout D3:** `dropoutHazard = 1/173.1234` (exponential)
- **Dropout D4:** `dropoutHazard = (1:20)/100` (piecewise exponential)
- **Dropout D5:** `dropoutShape = 1, dropoutScale = 173.1234` (Weibull, exponential)
- **Dropout D6:** user-defined function

Command-line Code

```
dsnD1 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropoutQuantiles=120, seed=0)
seqPlotPHNSubjects(dsnD1,theta=c(0.7,1))

dsnD2 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropoutQuantiles=c(120, 240), dropoutProbs=c(0.5, 0.75),
  seed=0)

dsnD3 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropoutHazard=1/173.1234, seed=0)

dsnD4 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropoutHazard=(1:20)/100, seed=0)
seqPlotPHNSubjects(dsnD4,theta=c(0.7,1))

dsnD5 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropoutShape=1, dropoutScale=173.1234, seed=0)

dropoutD6 <- function (n, dropoutArgs) {
  distn <- dropoutArgs$distn
  distn <- sort(distn)
  pdistn <- 1-c(0, (1:length(distn)) / length(distn))
  as.numeric(distn[cut(runif(n), 1-pdistn)])
}
```

```

}
set.seed(1)
dropoutArgsD6 <- list(distn=rlnorm(1000, 4, 1))
#descrip(dropoutArgs$distn)
#descrip(dropout(1000, dropoutArgs))
dsnD6 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropout=dropoutD6, dropoutArgs=dropoutArgsD6, seed=0)
seqPlotPHNSubjects(dsnD6, theta=c(0.7,1))

```

Output Output from example D6 is shown; the others are similar.

PROBABILITY MODEL and HYPOTHESES:

```

Two arm study of censored time to event response variable
Theta is hazard ratio (Treatment : Comparison)
One-sided hypothesis test of a lesser alternative:
  Null hypothesis : Theta >= 1      (size = 0.025)
  Alternative hypothesis : Theta <= 0.7  (power = 0.975)

```

STOPPING BOUNDARIES: Sample Mean scale

```

          a      d
Time 1 (NEv= 119.04) 0.4499 1.0872
Time 2 (NEv= 238.08) 0.6707 0.9557
Time 3 (NEv= 357.11) 0.7662 0.9026
Time 4 (NEv= 476.15) 0.8190 0.8724
Time 5 (NEv= 595.19) 0.8523 0.8523

```

ACCRUAL INFORMATION:

```

Accrual distribution:
  Uniform distribution
  Number of subjects:  Unspecified
  Accrual time       :    24
  Accrual rate       :  Unspecified
Event distribution:
  Exponential (hazard=0.05776227; 50th %ile=12)
Dropout distribution:
  User-specified function: dropoutD6

```

Accrual summary table:

	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	765	31.88	24	60
null	1.0	1	721	30.04	24	60

Timing of analyses:

```

Theta = 0.7  Scenario 1
          Analysis 1 Analysis 2 Analysis 3 Analysis 4 Analysis 5
Analysis Time    13.85    20.67    26.74    35.7     60.0
N Accrued       443.12    660.23    765.00    765.0    765.0
N Events        119.04    238.08    357.11    476.2    595.2

```

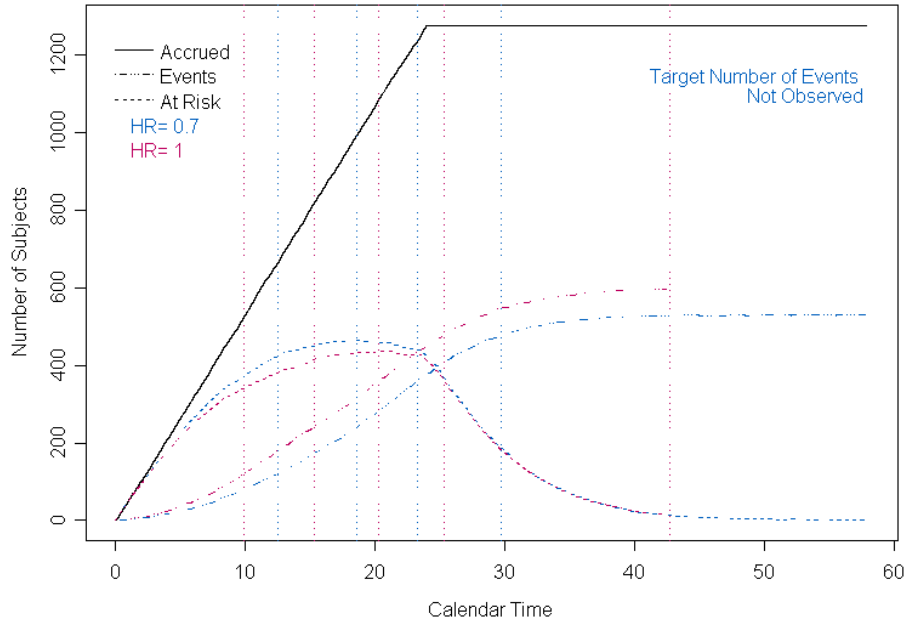
```

Theta = 1    Scenario 1

```


	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	13.2	19.99	25.88	34.24	60.0
N Accrued	398.1	600.51	721.00	721.00	721.0
N Events	119.0	238.08	357.11	476.15	595.2

Scenario 1 Designed for Theta 1



Example T1: Theta values for printout

This example provides a what to specify theta values (hazard ratios) are used in printouts of the Accrual summary and Timing of analyses tables. Examples A0 and T1 are identical, except for specification of the theta values for printouts.

Command-line Code

```
dsnA0 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualSize=700, accrualTime=24, seed=0,
  eventQuantiles=12)
```

```
dsnT1 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualSize=700, accrualTime=24, seed=0,
  eventQuantiles=12, accrTheta=(7:10)/10)
```

```
# Equivalent results can be obtained using seqPHNSubjects:
set.seed(0) # Use same random numbers as for dsnT1
seqPHNSubjects(dsnA0, accrTheta = 7:10/10)
```

Output The printout for A0 includes accrual summaries at the null and alternative hypotheses:

Accrual summary table:

	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
alternative	0.7	1	700	29.17	24	52.78
null	1.0	1	700	29.17	24	46.26

Timing of analyses:

Theta = 0.7	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	14.44	21.42	27.8	36.71	52.78
N Accrued	422.52	625.70	700.0	700.00	700.00
N Events	119.04	238.08	357.1	476.15	595.19

Theta = 1	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	13.39	20.02	25.67	33.05	46.26
N Accrued	391.60	584.48	700.00	700.00	700.00
N Events	119.04	238.08	357.11	476.15	595.19

The printout for T1, in contrast, includes the specified theta values:

Accrual summary table:

	theta	Scenario	Naccrual	accrualRate	accrualTime	studyTime
[1,]	0.7	1	700	29.17	24	52.78
[2,]	0.8	1	700	29.17	24	50.12
[3,]	0.9	1	700	29.17	24	48.00
[4,]	1.0	1	700	29.17	24	46.19

Timing of analyses:

Theta = 0.7	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	14.44	21.42	27.8	36.71	52.78
N Accrued	422.52	625.70	700.0	700.00	700.00
N Events	119.04	238.08	357.1	476.15	595.19

Theta = 0.8	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	14.04	20.96	26.97	35.24	50.12
N Accrued	410.86	612.07	700.00	700.00	700.00
N Events	119.04	238.08	357.11	476.15	595.19

Theta = 0.9	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	13.69	20.47	26.28	34.11	48.0
N Accrued	400.07	597.52	700.00	700.00	700.0
N Events	119.04	238.08	357.11	476.15	595.2

Theta = 1	Scenario 1				
	Analysis 1	Analysis 2	Analysis 3	Analysis 4	Analysis 5
Analysis Time	13.41	20.05	25.75	33.1	46.19
N Accrued	391.46	584.70	700.00	700.0	700.00
N Events	119.04	238.08	357.11	476.2	595.19

Example O2: Simulation-based Operating Characteristics

This example demonstrates the use of simulation-based operating characteristics. The accrual, event, and dropout specifications match earlier examples.

Summary:

- **Accrual:** Specify `accrualTime`, `studyTime`
uniform accrual rate, sample size is computer
- **Event:** `eventQuantiles = 12` (median for exponential distribution)
- **Dropout:** exponential with median 120
- **Operating Characteristics:** evaluate 7 alternatives

Command-line Code

```
dsn02 <- seqDesignExtd("hazard", nbr=5, test.type="less", alt=0.7,
  P=c(1, 0.5), accrualTime=24, studyTime=60, eventQuantiles=12,
  dropoutQuantiles=120, seed=0)
dsn02.OC <- seqOperatingCharExtd(dsn02,
  theta=c(1, 0.95, 0.9, 0.85, 0.8, 0.75, 0.7),
  Nsimul=1000)

dsn02.OC
plot(dsn02.OC)
seqPlotASN(list(AsympOC=dsn02.OC$AsympOC, SimulOC=dsn02.OC$SimulOC))
seqPlotPower(list(AsympOC=dsn02.OC$AsympOC, SimulOC=dsn02.OC$SimulOC))
seqPlotStopProb(list(AsympOC=dsn02.OC$AsympOC, SimulOC=dsn02.OC$SimulOC))
seqPlotInference(dsn02.OC)
```

Output

```
...
##Asymptotic Operating Characteristics
Operating characteristics
Theta      ASN Power.lower
1.00 260.9190    0.0250
0.95 309.8256    0.0853
0.90 362.1537    0.2267
0.85 400.7051    0.4606
0.80 406.9328    0.7186
0.75 378.8188    0.8985
0.70 333.6713    0.9750

Stopping Probabilities:
Theta Time 1 Time 2 Time 3 Time 4 Time 5
1.00 0.3241 0.3378 0.1999 0.0987 0.0396
0.95 0.2309 0.2839 0.2290 0.1642 0.0921
0.90 0.1513 0.2132 0.2388 0.2349 0.1617
0.85 0.0899 0.1551 0.2575 0.2938 0.2037
0.80 0.0480 0.1452 0.3201 0.3138 0.1729
0.75 0.0241 0.2139 0.4091 0.2615 0.0914
0.70 0.0161 0.3690 0.4379 0.1497 0.0273
```

##Simulated Operating Characteristics (Nsimul = 1000)

Operating characteristics

Theta	ASN	Power.lower
1.00	264.299	0.032
0.95	305.949	0.071
0.90	357.952	0.205
0.85	397.698	0.443
0.80	403.529	0.705
0.75	387.940	0.902
0.70	332.129	0.974

Stopping Probabilities:

Theta	Time 1	Time 2	Time 3	Time 4	Time 5
1.00	0.326	0.327	0.197	0.100	0.050
0.95	0.228	0.291	0.246	0.152	0.083
0.90	0.158	0.209	0.257	0.219	0.157
0.85	0.094	0.166	0.247	0.290	0.203
0.80	0.048	0.150	0.326	0.315	0.161
0.75	0.017	0.186	0.422	0.270	0.105
0.70	0.015	0.381	0.429	0.148	0.027

Mean Calendar Time at Analyses:

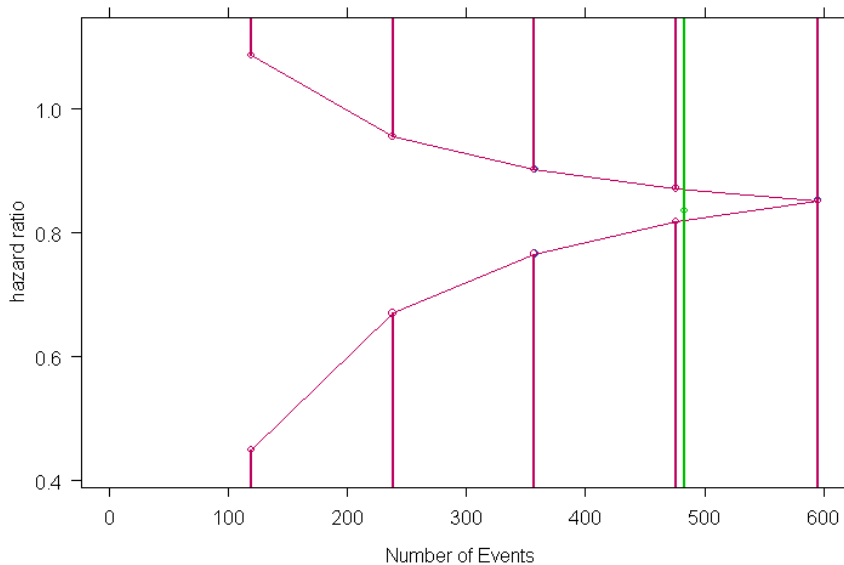
Theta	Time 1	Time 2	Time 3	Time 4	Time 5
1.00	13.3074	19.9616	25.7440	33.5963	50.0555
0.95	13.4726	20.1810	26.0716	34.1821	51.2402
0.90	13.6268	20.4002	26.3621	34.7055	52.5185
0.85	13.7625	20.6323	26.7433	35.4564	54.1693
0.80	13.9154	20.8816	27.1270	36.1008	55.6996
0.75	14.1209	21.1481	27.5320	36.8940	58.0324
0.70	14.3639	21.4532	28.0126	37.7999	60.2390

Mean Subjects Accrued at Analyses:

Theta	Time 1	Time 2	Time 3	Time 4	Time 5
1.00	402.936	604.307	725.974	726	726
0.95	407.925	611.034	725.998	726	726
0.90	412.485	617.586	726.000	726	726
0.85	416.612	624.338	726.000	726	726
0.80	420.953	632.228	726.000	726	726
0.75	428.284	640.210	726.000	726	726
0.70	434.029	649.157	726.000	726	726

Boundaries Used for Operating Characteristics

- — Boundary used for Asymptotics
- — Boundary used for Simulations
- — Fixed



Power

- Asymptotics
- 1000 Simulations

