

Package ‘MRTSampleSize’

July 25, 2017

Type Package

Title A Sample Size Calculator for Micro-Randomized Trials

Version 0.1.0

Depends R (>= 2.15.0)

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Description Provide a sample size calculator for micro-randomized trials (MRTs) based on methodology developed in Sample Size Calculations for Micro-randomized Trials in mHealth by Liao et al. (2016) <DOI:10.1002/sim.6847>.

License GPL (>= 2)

LazyData TRUE

RoxygenNote 5.0.1

NeedsCompilation no

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Repository CRAN

Date/Publication 2017-07-25 11:06:21 UTC

R topics documented:

calculatePower	2
calculateSampleSize	4
plotExpectAvail	6
plotProximalEffect	7

Index	9
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calculatePower	<i>Calculate power for micro-randomized trials</i>
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Description

This function calculates power for micro-randomized trials (MRTs) based on methodology developed in Sample Size Calculations for Micro-randomized Trials in mHealth by Liao et al. (2016) <DOI:10.1002/sim.6847>.

Usage

```
calculatePower(days, occ_per_day, prob, beta_shape, beta_mean, beta_initial,
  beta_quadratic_max, tau_shape, tau_mean, tau_initial, tau_quadratic_max,
  sample_size, sigLev)
```

Arguments

days	Duration of the study.
occ_per_day	Number of decision time points per day.
prob	Randomization probability, i.e. the probability of assigning the treatment at a decision time point. This can be constant, or time-varying probabilities can be specified by a vector specifying randomization probabilities for each day or decision time.
beta_shape	The trend for the proximal treatment effect, choices are constant, linear or quadratic. Note: <ol style="list-style-type: none"> 1. Constant The proximal treatment effect stays constant over the study. 2. Linear The linearly increasing form of a proximal treatment effect might be used if participants will get more enthusiastically engage in the apps and thus the proximal effect will increase as the study goes. The linearly decreasing form of a proximal treatment effect might be used if participants are likely to disengage the activity suggestions and thus the proximal effect will decrease as the study goes. 3. Quadratic The quadratic form of a proximal treatment effect might be used if you expect that initially participants will enthusiastically engage in the apps and thus the proximal effect will get higher. Then, as the study goes on, some participants are likely to disengage or begin to ignore the activity suggestions and hence a downward trend.
beta_mean	Average of proximal treatment effect.
beta_initial	Initial value of proximal treatment effect when beta_shape is linear or quadratic.
beta_quadratic_max	Day of maximal proximal treatment effect when beta_shape is quadratic.
tau_shape	The pattern for expected availability, choices are constant, linear or quadratic. Note: <ol style="list-style-type: none"> 1. Constant The expected availability stays constant over the study.

2. Linear A linearly increasing pattern of expected availability might be used if participants will find the intervention useful and thus more likely to turn the intervention on. A linearly decreasing pattern of expected availability might be used if participants learn more about the intervention and get bored through the course of the study and thus getting less likely to turn on the intervention.
3. Quadratic A quadratic pattern of availability. Here the changing point of availability refers to day of either maximal or minimal availability, depending on the input values of initial and average availability.

tau_mean	Average of expected availability.
tau_initial	Initial Value of expected availability when tau_shape is linear or quadratic.
tau_quadratic_max	Changing point of availability when tau_shape is quadratic.
sample_size	Number of participants
sigLev	Significance level

Value

power.

References

Seewald, N.J.; Sun, J.; Liao, P. "MRT-SS Calculator: An R Shiny Application for Sample Size Calculation in Micro-Randomized Trials". arXiv:1609.00695

Examples

```
calculatePower(days=42,
               occ_per_day=5,
               prob=0.4,
               beta_shape="quadratic",
               beta_mean=0.1,
               beta_initial=0,
               beta_quadratic_max=28,
               tau_shape="quadratic",
               tau_mean=0.5,
               tau_initial=0.7,
               tau_quadratic_max=42,
               sample_size=40,
               sigLev=0.05)

prob1 <- c(replicate(35,0.7),replicate(35,0.6),replicate(35,0.5),replicate(35,0.4))
calculatePower(days=28,
               occ_per_day=5,
               prob=prob1,
               beta_shape="quadratic",
               beta_mean=0.1,
               beta_initial=0,
               beta_quadratic_max=28,
```

```
tau_shape="quadratic",
tau_mean=0.5,
tau_initial=0.7,
tau_quadratic_max=42,
sample_size=40,
sigLev=0.05)#'
```

calculateSampleSize *Calculate sample size for micro-randomized trials*

Description

This function calculates the sample size for micro-randomized trials (MRTs) based on methodology developed in Sample Size Calculations for Micro-randomized Trials in mHealth by Liao et al. (2016) <DOI:10.1002/sim.6847>.

Usage

```
calculateSampleSize(days, occ_per_day, prob, beta_shape, beta_mean,
  beta_initial, beta_quadratic_max, tau_shape, tau_mean, tau_initial,
  tau_quadratic_max, power, sigLev)
```

Arguments

days	Duration of the study.
occ_per_day	Number of decision time points per day.
prob	Randomization probability, i.e. the probability of assigning the treatment at a decision time point. This can be constant, or time-varying probabilities can be specified by by a vector specifying randomization probabilities for each day or decision time.
beta_shape	The trend for the proximal treatment effect, choices are constant, linear or quadratic. Note: <ol style="list-style-type: none"> 1. Constant The proximal treatment effect stays constant over the study. 2. Linear The linearly increasing form of a proximal treatment effect might be used if participants will get more enthusiastically engage in the apps and thus the proximal effect will increase as the study goes. The linearly decreasing form of a proximal treatment effect might be used if participants are likely to disengage the activity suggestions and thus the proximal effect will decrease as the study goes. 3. Quadratic The quadratic form of a proximal treatment effect might be used if you expect that initially participants will enthusiastically engage in the apps and thus the proximal effect will get higher. Then, as the study goes on, some participants are likely to disengage or begin to ignore the activity suggestions and hence a downward trend.
beta_mean	Average of proximal treatment effect.

beta_initial	Initial value of proximal treatment effect when beta_shape is linear or quadratic.
beta_quadratic_max	Day of maximal proximal treatment effect when beta_shape is quadratic.
tau_shape	The pattern for expected availability, choices are constant, linear or quadratic. Note: <ol style="list-style-type: none"> 1. Constant The expected availability stays constant over the study. 2. Linear A linearly increasing pattern of expected availability might be used if participants will find the intervention useful and thus more likely to turn the intervention on. A linearly decreasing pattern of expected availability might be used if participants learn more about the intervention and get bored through the course of the study and thus getting less likely to turn on the intervention. 3. Quadratic A quadratic pattern of availability. Here the changing point of availability refers to day of either maximal or minimal availability, depending on the input values of initial and average availability.
tau_mean	Average of expected availability.
tau_initial	Initial Value of expected availability when tau_shape is linear or quadratic.
tau_quadratic_max	Changing point of availability when tau_shape is quadratic.
power	power
sigLev	Significance level

Value

Sample size.

References

Seewald, N.J.; Sun, J.; Liao, P. "MRT-SS Calculator: An R Shiny Application for Sample Size Calculation in Micro-Randomized Trials". arXiv:1609.00695

Examples

```
calculateSampleSize(days=42,
                    occ_per_day=5,
                    prob=0.4,
                    beta_shape="quadratic",
                    beta_mean=0.1,
                    beta_initial=0,
                    beta_quadratic_max=28,
                    tau_shape="quadratic",
                    tau_mean=0.5,
                    tau_initial=0.7,
                    tau_quadratic_max=42,
                    power=0.8,
                    sigLev=0.05)

prob1 <- c(replicate(35,0.7),replicate(35,0.6),replicate(35,0.5),replicate(35,0.4))
```

```

calculateSampleSize(days=28,
                    occ_per_day=5,
                    prob=prob1,
                    beta_shape="quadratic",
                    beta_mean=0.1,
                    beta_initial=0,
                    beta_quadratic_max=28,
                    tau_shape="quadratic",
                    tau_mean=0.5,
                    tau_initial=0.7,
                    tau_quadratic_max=42,
                    power=0.8,
                    sigLev=0.05)

```

plotExpectAvail *plot the graph for the expected availability*

Description

plot of the graphs for the expected availability, i.e., the expected probability that a participant is available to receive treatment at a decision time. when the pattern for the expected availability is constant, linear or quadratic.

Usage

```

plotExpectAvail(days, occ_per_day, tau_shape, tau_mean, tau_initial,
                tau_quadratic_max)

```

Arguments

days	Duration of the study.
occ_per_day	Number of decision time points per day.
tau_shape	The pattern for expected availability, choices are constant, linear or quadratic. Note: <ol style="list-style-type: none"> 1. Constant The expected availability stays constant over the study. 2. Linear A linearly increasing pattern of expected availability might be used if participants will find the intervention useful and thus more likely to turn the intervention on. A linearly decreasing pattern of expected availability might be used if participants learn more about the intervention and get bored through the course of the study and thus getting less likely to turn on the intervention. 3. Quadratic A quadratic pattern of availability. Here the changing point of availability refers to day of either maximal or minimal availability, depending on the input values of initial and average availability.
tau_mean	Average of expected availability.
tau_initial	Initial Value of expected availability when tau_shape is linear or quadratic.
tau_quadratic_max	Changing point of availability when tau_shape is quadratic.

Value

A graph for expected availability.

Examples

```
plotExpectAvail(days=42,
  occ_per_day=5,
  tau_shape="quadratic",
  tau_mean=0.5,
  tau_initial=0.7,
  tau_quadratic_max=42)
```

plotProximalEffect *plot the graph for the proximal treatment effect*

Description

plot of the graphs for the proximal treatment effect when the trend for the proximal treatment effect is constant, linear or quadratic.

Usage

```
plotProximalEffect(days, occ_per_day, beta_shape, beta_mean, beta_initial,
  beta_quadratic_max)
```

Arguments

days	Duration of the study.
occ_per_day	Number of decision time points per day.
beta_shape	The trend for the proximal treatment effect, choices are constant, linear or quadratic. Note: <ol style="list-style-type: none"> 1. Constant The proximal treatment effect stays constant over the study. 2. Linear The linearly increasing form of a proximal treatment effect might be used if participants will get more enthusiastically engage in the apps and thus the proximal effect will increase as the study goes. The linearly decreasing form of a proximal treatment effect might be used if participants are likely to disengage the activity suggestions and thus the proximal effect will decrease as the study goes. 3. Quadratic The quadratic form of a proximal treatment effect might be used if you expect that initially participants will enthusiastically engage in the apps and thus the proximal effect will get higher. Then, as the study goes on, some participants are likely to disengage or begin to ignore the activity suggestions and hence a downward trend.
beta_mean	Average of proximal treatment effect.

`beta_initial` Initial value of proximal treatment effect when `beta_shape` is linear or quadratic.
`beta_quadratic_max` Day of maximal proximal treatment effect when `beta_shape` is quadratic.

Value

A graph for the proximal treatment effect.

Examples

```
plotProximalEffect(days=42,  
                   occ_per_day=5,  
                   beta_shape="quadratic",  
                   beta_mean=0.1,  
                   beta_initial=0,  
                   beta_quadratic_max=28)
```


Index

calculatePower, [2](#)
calculateSampleSize, [4](#)

plotExpectAvail, [6](#)
plotProximalEffect, [7](#)