# Package 'swdpwr'

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Type Package
Title Power Calculation for Stepped Wedge Cluster Randomized Trials
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Description To meet the needs of statistical power calculation for stepped wedge cluster randomized trials, we developed this software. Different parameters can be specified by users for different scenarios, including: cross-sectional and cohort designs, binary and continuous outcomes, marginal (GEE) and conditional models (mixed effects model), three link functions (identity, log, logit links), with and without time effects under exchangeable, nested exchangeable and block exchangeable correlation structures. Unequal numbers of clusters per sequence are also allowed. The methods included in this package: Zhou et al. (2020) <doi:10.1093 biostatistics="" kxy031="">, Li et al. (2018) <doi:10.1111 biom.12918="">. Supplementary documents can be found at: <a href="https://ysph.yale.edu/cmips/research/software/swdpwr/">https://ysph.yale.edu/cmips/research/software/swdpwr/</a>. The Shiny app for swdpwr can be accessed at: <a href="https://jiachenchen322.shinyapps.io/swdpwr_shinyapp/">https://jiachenchen322.shinyapps.io/swdpwr_shinyapp/</a>. The package also includes functions that perform calculations for the intra-cluster correlation coefficients based on the random effects variances as input variables for continuous and binary outcomes, respectively.</doi:10.1111></doi:10.1093>
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swdpower	• • • • • • • •

# Description

This package includes a function swdpower that accounts for power calculation for stepped wedge cluster randomized trials.

#### **Details**

Package: swdpwr Type: Package Version: 1.5

Date: 2021-04-02 License: GPL (version 3)

Previous literature and developement of software focused mainly on continuous outcomes and obtained approximation results for binary outcomes. This package implemented new methods of power calculation for stepped wedge designs with binary outcomes and also incorporated procedures for continuous outcomes. The function swdpower can accommodate both cross-sectional and cohort designs, binary and continuous outcomes, marginal (GEE) and conditional models (mixed effects model), three link functions (identity, log, logit links), with and without time effects under exchangeable, nested exchangeable and block exchangeable correlation structures. Unequal numbers of clusters per sequence are also allowed. With this package, investigators can obtain more accurate calculation of statistical power, that will help a lot in the design and analysis of stepped wedge cluster randomized trials. The package also includes functions that perform calculations for the intra-cluster correlation coefficients based on the random effects variances as input variables for continuous and binary outcomes, respectively. Other supplementary documents can be found at: https://ysph.yale.edu/cmips/research/software/swdpwr/. The Shiny app for swdpwr can be accessed at: <a href="https://jiachenchen322.shinyapps.io/swdpwr\_shinyapp/">https://jiachenchen322.shinyapps.io/swdpwr\_shinyapp/</a>.

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#### References

Zhou X, Liao X, Kunz L M, et al. A maximum likelihood approach to power calculations for stepped wedge designs of binary outcomes[J]. Biostatistics, 2020, 21(1): 102-121.

Li F, Turner E L, Preisser J S. Sample size determination for GEE analyses of stepped wedge cluster randomized trials[J]. Biometrics, 2018, 74(4): 1450-1458.

BinICC

A function for calculating the intracluster correlation coefficient (ICC) for binary outcomes given the cluster level random effects variance

# Description

This function calculates the ICC (intracluster correlation coeffcient which measures the correlation between individuals in the same cluster) under different link funcitions in a cross-sectional stepped wedge CRT with binary outcomes. This model considers only the fixed time effects and does not include cluster by time interaction random effect.

#### Usage

```
BinICC(link = "identity", meanresponse_start, tau2 = 0)
```

#### **Arguments**

link choose link function from link="identity", link="log" and link="logit", with default value of identity link

meanresponse\_start

the anticipated mean response in the control group at the start of the study

tau2 also denoted as sigma\_b: variance of the between-cluster random effect, default

is 0

#### Value

The object returned includes the link function and value for the ICC in this study

#### **Examples**

```
BinICC(link="identity", meanresponse_start=0.2, tau2=0.05)
```

ContICC ContICC

ContICC	A function for calculating ICCs for continuous outcomes given random effects variances

# Description

This function calculates the within-period, between-period, and within-individual correlation parameters for continuous outcomes in a stepped wedge CRT

# Usage

```
ContICC(
  type = "cross-sectional",
  sigma2 = 1,
  sigma_b = 0,
  sigma_c = 0,
  sigma_pi = NA
)
```

# Arguments

type	choose the study type, specify type="cohort" for closed cohort study and type="cross-sectional" for cross-sectional study, default is "cross-sectional"
sigma2	marginal variance of the outcom, default is 1
sigma_b	variance of the between-cluster random effect, default is 0
sigma_c	variance of the cluster-by-time interaction random effect, default is 0
sigma_pi	variance of the random effect for repeated measures of one individual, this parameter should not be specified for cross-sectional studies, default is NA

# Value

The object returned includes the study type and values for the ICCs in this study

# **Examples**

```
ContICC(type="cohort", sigma2=1.5, sigma_b=0.5, sigma_c=0.2, sigma_pi=0.3)
```

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print.swdpower

Print the results of swdpower

#### **Description**

The print method for class "swdpower"

#### Usage

```
## S3 method for class 'swdpower' print(x, ...)
```

# Arguments

x an object used to select a method.

... further arguments passed to or from other methods.

#### Value

The output from print

swdpower

A function of power calculation for Stepped Wedge Design Studies

#### **Description**

This function performs power calculations for stepped wedge cluster randomized trials under different scenarios.

#### Usage

```
swdpower(
    K,
    design,
    family = "binomial",
    model = "conditional",
    link = "identity",
    type = "cross-sectional",
    meanresponse_start = NA,
    meanresponse_end0 = meanresponse_start,
    meanresponse_end1 = NA,
    effectsize_beta = NA,
    sigma2 = 0,
    typeIerror = 0.05,
    alpha0 = 0.1,
```

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```
alpha1 = alpha0/2,
alpha2 = NA
)
```

#### **Arguments**

K number of participants at each time period in a cluster, specified as the average

clusterperiod size considering cluster-size variability

design I\*J dimensional data set that describes the study design (control 0, intervention

1), I is the number of clusters, J is the number of time periods. Unequal allocation of sequences and only complete designs with no transition periods are

allowed

family family of responses, specify family="gaussian" for continuous outcome and

family="binomial" for binary outcome, with default value of "binomial"

model choose from conditional model (model="conditional") and marginal model (model="marginal"),

with default value of applying conditional model

link choose link function from link="identity", link="log" and link="logit", with de-

fault value of identity link

type choose the study type, specify type="cohort" for closed cohort study and type="cross-

sectional" for cross-sectional study, with default value of cross-sectional study

meanresponse\_start

the anticipated mean response in the control group at the start of the study

meanresponse\_end0

the anticipated mean response in the control group at the end of the study, with

default value equals to meanresponse\_start (no time effects)

meanresponse\_end1

the anticipated mean response in the intervention group at the end of the study

effectsize beta

the anticipated effect size, just omit this parameter if you don't need to specify it. In all scenarios, you can choose to specify the three parameters about mean responses without specifying this effect size, or alternatively specify meanresponse\_start, meanresponse\_end0 and this effect size. For continuous outcomes, users can conduct power calculations by only specifying this parameter without the above three parameters about mean responses (as the power is dependent just on it), then calculation will be implemented assuming scenarios without time effects. If you would consider scenarios with time effects and continuous outcomes, please specify meanresponse\_start, meanresponse\_end0 (donot require accurate information, just make sure they are not equal) and this effectsize\_beta.

sigma2 marginal variance of the outcome (only needed for continuous outcomes and

should not be an input for binary outcomes), with default value of 0.

typeIerror two-sided type I error, with default value of 0.05 alpha0 within-period correlation, with default value of 0.1

alpha1 between-period correlation, with default value of alpha0/2

alpha2 within-individual correlation, should not be an input under cross-sectional de-

signs although it is numerically identical to alpha1 in this scenario by definition

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#### Value

The object returned has a class of swdpower, which includes a list of the design matrix and a summary of this design (including the power)

#### **Examples**

```
library(swdpwr)
#a cross-sectional design with 12 clusters, 3 periods and binary outcomes applying conditional model
#alpha2 should not be specified, as the current version does not support power calculation using
#conditional models with binary outcomes in a cohort design
#create a 12*3 matrix which describes the study design,
#0 means control status, 1 means intervention status
dataset = matrix(c(rep(c(0,1,1),6),rep(c(0,0,1),6)),12,3,byrow=TRUE)
#specify meanresponse_start, meanresponse_end0 and meanresponse_end1
swdpower(K = 30, design = dataset, family = "binomial", model = "conditional", link = "logit",
type = "cross-sectional", meanresponse_start = 0.2, meanresponse_end0 = 0.3,
meanresponse_end1 = 0.4, typeIerror = 0.05, alpha0 = 0.01, alpha1 = 0.01)
#specify meanresponse_start, meanresponse_end0 and effectsize_beta
swdpower(K = 30, design = dataset, family = "binomial", model = "conditional", link = "logit",
type = "cross-sectional", meanresponse_start = 0.2, meanresponse_end0 = 0.3, effectsize_beta = 0.6,
typeIerror = 0.05, alpha0 = 0.01, alpha1 = 0.01)
#a cohort design with 8 clusters, 3 periods and continuous outcomes applying marginal model
#sigma2 should be specified, as continuous outcomes require marginal variance in calculation
#create a 8*3 matrix which describes the study design,
#0 means control status, 1 means intervention status
dataset = matrix(c(rep(c(0,1,1),4),rep(c(0,0,1),4)),8,3, byrow=TRUE)
#specify meanresponse_start, meanresponse_end0 and meanresponse_end1
swdpower(K = 24, design = dataset, family = "gaussian", model = "marginal", link = "identity",
type = "cohort", meanresponse_start = 0.1, meanresponse_end0 = 0.2, meanresponse_end1 = 0.4,
sigma2 = 0.095, typeIerror = 0.05, alpha0 = 0.03, alpha1 = 0.015, alpha2 = 0.2)
#specify effectsize_beta only, then the program runs assuming no time effects
swdpower(K = 24, design = dataset, family = "gaussian", model = "marginal", link = "identity",
type = "cohort",effectsize_beta=0.3, sigma2 = 0.095, typeIerror = 0.05, alpha0 = 0.03,
alpha1 = 0.015, alpha2 = 0.2)
```

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