

# Package ‘BDEsize’

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**Title** Efficient Determination of Sample Size in Balanced Design of Experiments

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**Description** For a balanced design of experiments, this package calculates the sample size required to detect a certain standardized effect size, under a significance level. This package also provides three graphs; detectable standardized effect size vs power, sample size vs detectable standardized effect size, and sample size vs power, which show the mutual relationship between the sample size, power and the detectable standardized effect size. The detailed procedure is described in R. V. Lenth (2006-9) <<https://homepage.divms.uiowa.edu/~rlenth/Power/>>, Y. B. Lim (1998), M. A. Kastenbaum, D. G. Hoel and K. O. Bowman (1970) <doi:10.2307/2334851>, and Douglas C. Montgomery (2013, ISBN: 0849323312).

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plots.2levFr	<i>Diagnosis Graphs for Sample Size of Two-level Fractional Factorial Design</i>
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### Description

This function produces graphs between the sample size, power and the detectable standardized effect size of two-level fractional factorial design.

### Usage

```
plots.2levFr(nfactor, nfraction, interaction = FALSE, delta_type = 1,
             delta = c(1, 0, 1), deltao = NULL, alpha = 0.05, beta = 0.2, type = 1,
             maxsize = 1000)
```

### Arguments

nfactor	the number of factor.
nfraction	the number of fraction. For example, when a model is $2^{(k-p)}$ , k is the number of factor and p is the number of fraction. It is called a $1/2^p$ fraction of the $2^k$ design.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When <code>interaction = TRUE</code> , two-way interaction effects are include in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: <code>delta[1]</code> for main effects, <code>delta[2]</code> for two-way interaction effects, and <code>delta[3]</code> for standard deviation of noise. When <code>interaction=FALSE</code> , <code>delta[2]</code> is 0.
deltao	the minimal detectable standardized effect size for power vs the sample size plot when <code>type = 3</code> .
alpha	Type I error.
beta	Type II error.
type	graph type: 1 for Power vs Delta plot, 2 for Delta vs Sample size plot, and 3 for Power vs Sample size plot.
maxsize	tolerance for sample size.

**Details**

This function produces graph between the sample size, power 1-beta and the detectable standardized effect size delta of two-level fractional factorial design. According to type, it displays plot of Power vs Delta, Delta vs Sample size, or Power vs Sample size.

**Value**

plot of Power vs Delta, Delta vs Sample size, or Power vs Sample size according to type.

**See Also**

[plots.Full](#), [plots.Split](#), [plots.Block](#).

**Examples**

```
# plot of Power vs Delta for two-level fractional factorial design
# without the interaction effects
plots.2levFr(nfactor=3, nfraction=1, interaction=FALSE,
  delta_type=1, delta=c(1, 0, 1), alpha=0.05, beta=0.2, type=1)
# plot of Power vs Sample size for two-level fractional factorial design
# with the interaction effects
plots.2levFr(nfactor=5, nfraction=1, interaction=TRUE,
  delta_type=1, delta=c(1, 1, 1), deltao=1, alpha=0.05, beta=0.2, type=3)
```

---

plots.Block

*Diagnosis Graphs for the number of Blocks of Randomized Complete Block Design*

---

**Description**

This function produces graphs between the sample size, power and the detectable standardized effect size of randomized complete block design.

**Usage**

```
plots.Block(factor.lev, interaction = FALSE, delta_type = 1, delta = c(1, 0, 1),
  deltao = NULL, alpha = 0.05, beta = 0.2, type = 1, maxsize = 1000)
```

**Arguments**

factor.lev	vector of the numbers of levels for each factor.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When interaction = TRUE, two-way interaction effects are include in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.

delta	vector of effect sizes: delta[1] for main effects, delta[2] for two-way interaction effects, and delta[3] for standard deviation of noise. When interaction=FALSE, delta[2] is 0.
deltao	the minimal detectable standardized effect size for power vs the number of blocks plot when type = 3.
alpha	Type I error.
beta	Type II error.
type	graph type: 1 for Power vs Delta plot, 2 for Delta vs the Number of Blocks plot, and 3 for Power vs the Number of Blocks plot.
maxsize	tolerance for the number of blocks.

### Details

In a randomized complete block design (without replications), the optimal number of blocks need to be determined. This function produces graph between Number of Block, power 1-beta and the detectable standardized effect size delta of randomized complete block design. According to type, it displays plot of Power vs Delta, Delta vs Number of Blocks, or Power vs Number of Blocks.

### Value

plot of Power vs Delta, Delta vs Number of Blocks, or Power vs Number of Blocks according to type.

### See Also

[plots.Full](#), [plots.2levFr](#), [plots.Split](#).

### Examples

```
# plot of Power vs Delta for randomized complete block design
# with 2 factors without the interaction effects
plots.Block(factor.lev=c(2, 2), interaction=FALSE,
            delta_type=1, delta=c(1, 0, 1), alpha=0.05, beta=0.2, type=1)

# plot of Power vs Number of Blocks for randomized complete block design
# with 2 factors with the interaction effects
plots.Block(factor.lev=c(2, 3), interaction=TRUE,
            delta_type=1, delta=c(1, 1, 1), deltao=1.5, alpha=0.05, beta=0.2, type=3)
```

---

plots.Full

*Diagnosis Graphs for Sample Size of Full Factorial Design*

---

### Description

This function produces graphs between the sample size, power and the detectable standardized effect size of full factorial design.

**Usage**

```
plots.Full(factor.lev, interaction = FALSE, delta_type = 1, delta = c(1, 0, 1),
           deltax = NULL, alpha = 0.05, beta = 0.2, type = 1, maxsize = 1000)
```

**Arguments**

factor.lev	vector of the numbers of levels for each factor.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When <code>interaction = TRUE</code> , two-way interaction effects are included in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: <code>delta[1]</code> for main effects, <code>delta[2]</code> for two-way interaction effects, and <code>delta[3]</code> for standard deviation of noise. When <code>interaction=FALSE</code> , <code>delta[2]</code> is 0.
deltax	the minimal detectable standardized effect size for power vs the sample size plot when <code>type = 3</code> .
alpha	Type I error.
beta	Type II error.
type	graph type: 1 for Power vs Delta plot, 2 for Delta vs Sample size plot, and 3 for Power vs Sample size plot.
maxsize	tolerance for sample size.

**Details**

This function produces graph between the sample size, power  $1-\beta$  and the detectable standardized effect size `delta` of full factorial design. According to `type`, it displays plot of Power vs Delta, Delta vs Sample size, or Power vs Sample size.

**Value**

plot of Power vs Delta, Delta vs Sample size, or Power vs Sample size according to `type`.

**See Also**

[plots.2levFr](#), [plots.Split](#), [plots.Block](#).

**Examples**

```
# plot of Power vs Delta for full factorial design
# with 2 factors without the interaction effects
plots.Full(factor.lev=c(2, 3), interaction=FALSE,
           delta_type=1, delta=c(1, 0, 1), alpha=0.05, beta=0.2, type=1)

# plot of Power vs Sample size for full factorial design
# with 2 factors with the interaction effects
plots.Full(factor.lev=c(2, 3), interaction=TRUE,
           delta_type=1, delta=c(1, 1, 1), deltax=1.5, alpha=0.05, beta=0.2, type=3)
```

plots.Split

*Diagnosis Graphs for Sample Size of Split-Plot Design***Description**

This function produces graphs between the sample size, power and the detectable standardized effect size of split-plot design.

**Usage**

```
plots.Split(whole.factor.lev, split.factor.lev, interaction = FALSE,
            delta_type = 1, delta = c(1, 0, 1, 1), del tao = NULL, alpha = 0.05, beta = 0.2,
            type = 1, maxsize = 1000)
```

**Arguments**

whole.factor.lev	vector of the numbers of levels for each whole factor.
split.factor.lev	vector of the numbers of levels for each split factor.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When interaction = TRUE, two-way interaction effects are include in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: delta[1] for main effects, delta[2] for two-way interaction effects, and delta[3] and delta[4] for standard deviation of whole-plot noise and subplot noise, respectively. When interaction=FALSE, delta[2] is 0.
del tao	the minimal detectable standardized effect size for power vs the sample size plot when type = 3.
alpha	Type I error.
beta	Type II error.
type	graph type: 1 for Power vs Delta plot, 2 for Delta vs Sample size plot, and 3 for Power vs Sample size plot.
maxsize	tolerance for sample size.

**Details**

This function produces graph between the sample size, power  $1-\beta$  and the detectable standardized effect size  $\delta$  of split-plot design. According to type, it displays plot of Power vs Delta, Delta vs Sample size, or Power vs Sample size. The number of whole-plot factors and split plot factors are up to 2 in the current package version.

**Value**

plot of Power vs Delta, Delta vs Sample size, or Power vs Sample size according to type.

**See Also**

[plots.Full](#), [plots.2levFr](#), [plots.Block](#).

**Examples**

```
# plot of Power vs Delta for split-plot design
# without the interaction effects
plots.Split(whole.factor.lev=2, split.factor.lev=2, interaction=FALSE,
            delta_type=1, delta=c(1, 0, 1, 1), alpha=0.05, beta=0.2, type=1)

# plot of Power vs Sample size for split-plot design
# with the interaction effects
plots.Split(whole.factor.lev=2, split.factor.lev=2, interaction=TRUE,
            delta_type=1, delta=c(1, 1, 1, 1), deltao=1, alpha=0.05, beta=0.2, type=3)
```

---

Size.2levFr

*Sample Size Calculator for Two-level Fractional Factorial Design*

---

**Description**

This function computes sample size for two-level fractional factorial design to detect a certain standardized effect size with power at the significance level. The model for fractional factorial design contains only main effects in resolution III and IV.

**Usage**

```
Size.2levFr(nfactor, nfraction, interaction = FALSE, delta_type = 1,
            delta = c(1, 0, 1), alpha = 0.05, beta = 0.2, maxsize = 1000)
```

**Arguments**

nfactor	the number of factor.
nfraction	the number of fraction. For example, when a model is $2^{(k-p)}$ , k is the number of factor and p is the number of fraction. It is called a $1/2^p$ fraction of the $2^k$ design.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When interaction = TRUE, two-way interaction effects are include in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: delta[1] for main effects, delta[2] for two-way interaction effects, and delta[3] for standard deviation of noise. When interaction=FALSE, delta[2] is 0.

alpha	Type I error.
beta	Type II error.
maxsize	tolerance for sample size.

### Details

This function computes sample size in two-level fractional factorial design to detect a certain standardized effect size  $\delta$  with power  $1-\beta$  at the significance level  $\alpha$ .

### Value

model	a character vector expressing a model. The main effects are expressed by the upper-case letters of the Roman alphabet, and two-way interaction effects are denoted by * operator for pairs of the main effects.
n	optimal sample size.
Delta	a vector of minimal detectable standardized effect sizes.

### References

- R. V. Lenth (2006-9). Java Applets for Power and Sample Size[Computer software]. Retrieved March 27, 2018 from <https://homepage.divms.uiowa.edu/~rlenth/Power/>.
- Y. B. Lim (1998). Study on the Size of Minimal Standardized Detectable Difference in Balanced Design of Experiments. *Journal of the Korean society for Quality Management*, **26(4)**, 239–249.
- M. A. Kastenbaum, D. G. Hoel and K. O. Bowman (1970) Sample size requirements : one-way analysis of variance, *Biometrika*, **57(2)**, 421–430.
- D. C. Montgomery (2013) Design and analysis of experiments. John Wiley & Sons.

### See Also

[Size.Full](#), [Size.Split](#), [Size.Block](#).

### Examples

```
# only main effects
model1 <- Size.2levFr(nfactor=3, nfraction=1, interaction=FALSE,
  delta_type=1, delta=c(1, 0, 1), alpha=0.05, beta=0.2)
model1$model
model1$n
model1$Delta

# including two-way interaction effects
model2 <- Size.2levFr(nfactor=5, nfraction=1, interaction=TRUE,
  delta_type=1, delta=c(1, 1, 1), alpha=0.05, beta=0.2)
```



---

Size.Block	<i>The number of Blocks Calculator for Randomized Complete Block Design</i>
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---

### Description

This function computes the number of blocks for randomized complete block design to detect a certain standardized effect size with power at the significance level.

### Usage

```
Size.Block(factor.lev, interaction = FALSE, delta_type = 1, delta = c(1, 0, 1),
           alpha = 0.05, beta = 0.2, maxsize = 1000)
```

### Arguments

factor.lev	vector of the numbers of levels for each factor.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When <code>interaction = TRUE</code> , two-way interaction effects are included in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: <code>delta[1]</code> for main effects, <code>delta[2]</code> for two-way interaction effects, and <code>delta[3]</code> for standard deviation of noise. When <code>interaction=FALSE</code> , <code>delta[2]</code> is 0.
alpha	Type I error.
beta	Type II error.
maxsize	tolerance for the number of blocks.

### Details

In a randomized complete block design (without replications), the optimal number of blocks need to be determined. This function computes the number of blocks for randomized complete block design to detect a certain standardized effect size `delta` with power  $1-\beta$  at the significance level `alpha`.

### Value

model	a character vector expressing a model. The main effects are expressed by the upper-case letters of the Roman alphabet, and two-way interaction effects are denoted by <code>*</code> operator for pairs of the main effects. The block factor is denoted by <code>Block</code> .
n	optimal the number of blocks.
Delta	a vector of minimal detectable standardized effect sizes.

## References

- R. V. Lenth (2006-9). Java Applets for Power and Sample Size[Computer software]. Retrieved March 27, 2018 from <https://homepage.divms.uiowa.edu/~rlenth/Power/>.
- Y. B. Lim (1998). Study on the Size of Minimal Standardized Detectable Difference in Balanced Design of Experiments. *Journal of the Korean society for Quality Management*, **26(4)**, 239–249.
- M. A. Kastenbaum, D. G. Hoel and K. O. Bowman (1970) Sample size requirements : one-way analysis of variance, *Biometrika*, **57(2)**, 421–430.
- D. C. Montgomery (2013) Design and analysis of experiments. John Wiley & Sons.

## See Also

[Size.Full](#), [Size.2levFr](#), [Size.Split](#).

## Examples

```
# only main effects
model1 <- Size.Block(factor.lev=c(2, 2), interaction=FALSE,
  delta_type=1, delta=c(1, 0, 1), alpha=0.05, beta=0.2)
model1$model
model1$n
model1$Delta

# including two-way interaction effects
model2 <- Size.Block(factor.lev=c(2, 2), interaction=TRUE,
  delta_type=1, delta=c(1, 1, 1), alpha=0.05, beta=0.2)
model2
```

---

Size.Full

*Sample Size Calculator for Full Factorial Design*

---

## Description

This function computes sample size for full factorial design to detect a certain standardized effect size with power at the significance level.

## Usage

```
Size.Full(factor.lev, interaction = FALSE, delta_type = 1, delta = c(1, 0, 1),
  alpha = 0.05, beta = 0.2, maxsize = 1000)
```

## Arguments

`factor.lev` vector of the numbers of levels for each factor.

`interaction` specifies whether two-way interaction effects are included in a model with the main effects. When `interaction = TRUE`, two-way interaction effects are included in a model.

delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: delta[1] for main effects, delta[2] for two-way interaction effects, and delta[3] for standard deviation of noise. When interaction=FALSE, delta[2] is 0.
alpha	Type I error.
beta	Type II error.
maxsize	tolerance for sample size.

### Details

This function computes sample size in full factorial design to detect a certain standardized effect size delta with power 1-beta at the significance level alpha.

### Value

model	a character vector expressing a model. The main effects are expressed by the upper-case letters of the Roman alphabet, and two-way interaction effects are denoted by * operator for pairs of the main effects.
n	optimal sample size.
Delta	a vector of minimal detectable standardized effect sizes.

### References

- R. V. Lenth (2006-9). Java Applets for Power and Sample Size[Computer software]. Retrieved March 27, 2018 from <https://homepage.divms.uiowa.edu/~rlenth/Power/>.
- Y. B. Lim (1998). Study on the Size of Minimal Standardized Detectable Difference in Balanced Design of Experiments. *Journal of the Korean society for Quality Management*, **26(4)**, 239–249.
- M. A. Kastenbaum, D. G. Hoel and K. O. Bowman (1970) Sample size requirements : one-way analysis of variance, *Biometrika*, **57(2)**, 421–430.
- D. C. Montgomery (2013) Design and analysis of experiments. John Wiley & Sons.

### See Also

[Size.2levFr](#), [Size.Split](#), [Size.Block](#).

### Examples

```
# only main effects
model1 <- Size.Full(factor.lev=c(2, 2), interaction=FALSE,
  delta_type=1, delta=c(1, 0, 1), alpha=0.05, beta=0.2)
model1$model
model1$n
model1$Delta

# including two-way interaction effects
model2 <- Size.Full(factor.lev=c(2, 2), interaction=TRUE,
  delta_type=1, delta=c(1, 1, 1), alpha=0.05, beta=0.2)
model2
```

Size.Split

*Sample Size Calculator for Split-Plot Design***Description**

This function computes sample size for split-plot design to detect a certain standardized effect size with power at the significance level.

**Usage**

```
Size.Split(whole.factor.lev, split.factor.lev, interaction = FALSE,
           delta_type = 1, delta = c(1, 0, 1, 1), alpha = 0.05, beta = 0.2,
           maxsize = 1000)
```

**Arguments**

whole.factor.lev	vector of the numbers of levels for each whole factor.
split.factor.lev	vector of the numbers of levels for each split factor.
interaction	specifies whether two-way interaction effects are included in a model with the main effects. When <code>interaction = TRUE</code> , two-way interaction effects are included in a model.
delta_type	specifies the type of standardized effect size: 1 for standard deviation type and 2 for range type.
delta	vector of effect sizes: <code>delta[1]</code> for main effects, <code>delta[2]</code> for two-way interaction effects, and <code>delta[3]</code> and <code>delta[4]</code> for standard deviation of whole-plot noise and subplot noise, respectively. When <code>interaction=FALSE</code> , <code>delta[2]</code> is 0.
alpha	Type I error.
beta	Type II error.
maxsize	tolerance for sample size.

**Details**

This function computes sample size in split-plot design to detect a certain standardized effect size  $\delta$  with power  $1-\beta$  at the significance level  $\alpha$ . The number of whole-plot factors and split plot factors are up to 2 in the current package version. The linear model for the split-plot design is

$$y_{ijklm} = \mu + \tau_i + \beta_j + \gamma_k + (\beta\tau)_{ik} + \theta_{ijk} + \delta_l + \lambda_m + (\delta\lambda)_{im} + (\beta\delta)_{jl} + (\beta\lambda)_{jm} + (\gamma\delta)_{kl} + (\delta\lambda)_{lm} + \epsilon_{ijklm}$$

where  $\tau_i$  is the replicate effect,  $\beta_j, \gamma_k$  is the whole-plot main effects,  $\theta_{ijk}$  is the whole-plot error,  $\delta_l, \lambda_m$  is the subplot main effects, and  $\epsilon_{ijklm}$  is the subplot error.

**Value**

model	a character vector expressing a model. The whole factor effects and the split factor effects are expressed by the lower-case letters and sequential upper-case letters of the Roman alphabet, and two-way interaction effects are denoted by * operator for pairs of the those effects.
n	optimal sample size.
Delta	a vector of minimal detectable standardized effect sizes.

**References**

- R. V. Lenth (2006-9). Java Applets for Power and Sample Size[Computer software]. Retrieved March 27, 2018 from <https://homepage.divms.uiowa.edu/~rlenth/Power/>.
- Y. B. Lim (1998). Study on the Size of Minimal Standardized Detectable Difference in Balanced Design of Experiments. *Journal of the Korean society for Quality Management*, **26(4)**, 239–249.
- M. A. Kastenbaum, D. G. Hoel and K. O. Bowman (1970) Sample size requirements : one-way analysis of variance, *Biometrika*, **57(2)**, 421–430.
- D. C. Montgomery (2013) Design and analysis of experiments. John Wiley & Sons.

**See Also**

[Size.Full](#), [Size.2levFr](#), [Size.Block](#).

**Examples**

```
# only main effects
splitmodel1 <- Size.Split(whole.factor.lev=c(2, 2), split.factor.lev=c(2, 2), interaction=FALSE,
  delta_type=1, delta=c(1, 0, 1, 1), alpha=0.05, beta=0.2)
splitmodel1$model
splitmodel1$n
splitmodel1$Delta

# including two-way interaction effects
splitmodel2 <- Size.Split(whole.factor.lev=c(2, 2), split.factor.lev=c(2, 2), interaction=TRUE,
  delta_type=1, delta=c(1, 1, 1, 1), alpha=0.05, beta=0.2)
splitmodel2
```

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