

# Package ‘ri2’

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**Type** Package

**Title** Randomization Inference for Randomized Experiments

**Version** 0.4.0

**Description** Randomization inference procedures for simple and complex randomized designs, including multi-armed trials, as described in Gerber and Green (2012, ISBN: 978-0393979954). Users formally describe their randomization procedure and test statistic. The randomization distribution of the test statistic under some null hypothesis is efficiently simulated.

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**Encoding** UTF-8

**Imports** generics, ggplot2, pbapply

**Depends** randomizr (>= 0.16.0), estimatr

**Suggests** testthat, knitr, rmarkdown

**RoxygenNote** 7.1.2

**VignetteBuilder** knitr

**NeedsCompilation** no

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**Repository** CRAN

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|------------|--|
| conduct_ri | <i>Conduct Randomization Inference</i> |
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## Description

This function makes it easy to conduct three kinds of randomization inference.

## Usage

```
conduct_ri(
  formula = NULL,
  model_1 = NULL,
  model_2 = NULL,
  test_function = NULL,
  assignment = "Z",
  outcome = NULL,
  declaration = NULL,
  sharp_hypothesis = 0,
  studentize = FALSE,
  IPW = TRUE,
  IPW_weights = NULL,
  sampling_weights = NULL,
  permutation_matrix = NULL,
  data,
  sims = 1000,
  progress_bar = FALSE,
  p = "two-tailed"
)
```

## Arguments

|               |  |
|---------------|--|
| formula       | an object of class formula, as in <a href="#">lm</a> . Use formula when conducting significance tests of an Average Treatment Effect estimate under a sharp null hypothesis. For the difference-in-means estimate, do not include covariates. For the OLS covariate-adjusted estimate, include covariates. |
| model_1       | an object of class formula, as in <a href="#">lm</a> . Models 1 and 2 must be "nested." model_1 should be the "restricted" model and model_2 should be the "unrestricted" model.   |
| model_2       | an object of class formula, as in <a href="#">lm</a> . Models 1 and 2 must be "nested." model_1 should be the "restricted" model and model_2 should be the "unrestricted" model.   |
| test_function | A function that takes data and returns a scalar test statistic.  |
| assignment    | a character string that indicates which variable is randomly assigned. Defaults to "Z".  |
| outcome       | a character string that indicates which variable is the outcome variable. Defaults to NULL.  |
| declaration   | A random assignment declaration, created by <a href="#">declare_ra</a> .   |

|                    |   |
|--------------------|---|
| sharp_hypothesis   | either a numeric scalar or a numeric vector of length $k - 1$ , where $k$ is the number of treatment conditions. In a two-arm trial, this number is the *hypothesized* difference between the treated and untreated potential outcomes for each unit.. In a multi-arm trial, each number in the vector is the hypothesized difference in potential outcomes between the baseline condition and each successive treatment condition. |
| studentize         | logical, defaults to FALSE. Should the test statistic be the t-ratio rather than the estimated ATE? T-ratios will be calculated using HC2 robust standard errors or their clustered equivalent. CLUSTERING NOT YET IMPLEMENTED.   |
| IPW                | logical, defaults to TRUE. Should inverse probability weights be calculated?  |
| IPW_weights        | a character string that indicates which variable is the existing inverse probability weights vector. Usually unnecessary, as IPW weights will be incorporated automatically if IPW = TRUE. Defaults to NULL.  |
| sampling_weights   | a character string that indicates which variable is the sampling weights vector. Optional, defaults to NULL. NOT YET IMPLEMENTED  |
| permutation_matrix | An optional matrix of random assignments, typically created by <a href="#">obtain_permutation_matrix</a> .  |
| data               | A data.frame.   |
| sims               | the number of simulations. Defaults to 1000.  |
| progress_bar       | logical, defaults to FALSE. Should a progress bar be displayed in the console?  |
| p                  | Should "two-tailed", "upper", or "lower" p-values be reported? Defaults to "two-tailed". For two-tailed p-values, whether or not a simulated value is as large or larger than the observed value is determined with respect to the distance to the sharp null.  |

## Details

1. Conduct hypothesis tests under the sharp null when the test statistic is the difference-in-means or covariate-adjusted average treatment effect estimate. 2. Conduct "ANOVA" style hypothesis tests, where the f-statistic from two nested models is the test statistic. This procedure is especially helpful when testing interaction terms under null of constant effects. 3. Arbitrary (scalar) test statistics

## Examples

```
# Data from Gerber and Green Table 2.2

# Randomization Inference for the Average Treatment Effect

table_2.2 <-
  data.frame(d = c(1, 0, 0, 0, 0, 0, 1),
            y = c(15, 15, 20, 20, 10, 15, 30))

## Declare randomization procedure
declaration <- declare_ra(N = 7, m = 2)
```

```

## Conduct Randomization Inference
out <- conduct_ri(y ~ d,
  declaration = declaration,
  assignment = "d",
  sharp_hypothesis = 0,
  data = table_2.2)

summary(out)
plot(out)
tidy(out)

# Using a custom permutation matrix

permutation_matrix <-
  matrix(c(0, 0, 0, 0, 0, 0, 1,
           0, 0, 0, 0, 0, 1, 0,
           0, 0, 0, 0, 1, 0, 0,
           0, 0, 0, 1, 0, 0, 0,
           0, 0, 1, 0, 0, 0, 0,
           0, 1, 0, 0, 0, 0, 0,
           1, 0, 0, 0, 0, 0, 0),
    ncol = 7)

conduct_ri(y ~ d, assignment = "d", data = table_2.2,
  permutation_matrix = permutation_matrix)

# Randomization Inference for an Interaction

N <- 100
declaration <- randomizr::declare_ra(N = N, m = 50)

Z <- randomizr::conduct_ra(declaration)
X <- rnorm(N)
Y <- .9 * X + .2 * Z + 1 * X * Z + rnorm(N)
dat <- data.frame(Y, X, Z)

ate_obs <- coef(lm(Y ~ Z, data = dat))[2]

out <-
  conduct_ri(
    model_1 = Y ~ Z + X,
    model_2 = Y ~ Z + X + Z * X,
    declaration = declaration,
    assignment = "Z",
    sharp_hypothesis = ate_obs,
    data = dat, sims = 100
  )

plot(out)

```

```
summary(out)

summary(out, p = "two-tailed")
summary(out, p = "upper")
summary(out, p = "lower")

tidy(out)

# Randomization Inference for arbitrary test statistics

## In this example we're conducting a randomization check (in this case, a balance test).

N <- 100
declaration <- randomizr::declare_ra(N = N, m = 50)

Z <- randomizr::conduct_ra(declaration)
X <- rnorm(N)
Y <- .9 * X + .2 * Z + rnorm(N)
dat <- data.frame(Y, X, Z)

balance_fun <- function(data) {
  f_stat <- summary(lm(Z ~ X, data = data))$f[1]
  names(f_stat) <- NULL
  return(f_stat)
}

## confirm function works as expected
balance_fun(dat)

## conduct randomization inference

out <-
  conduct_ri(
    test_function = balance_fun,
    declaration = declaration,
    assignment = "Z",
    sharp_hypothesis = 0,
    data = dat, sims = 100
  )

plot(out)
summary(out)
tidy(out)
```

## Description

Randomization Inference

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