

# Package ‘didimputation’

March 9, 2026

**Type** Package

**Title** Imputation Estimator from Borusyak, Jaravel, and Spiess (2021)

**Version** 0.5.1

**Description** Estimates Two-way Fixed Effects difference-in-differences/event-study models using the imputation-based approach proposed by Borusyak, Jaravel, and Spiess (2021).

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.3.2

**Depends** R (>= 4.1.0), fixest (>= 0.13.2), data.table (>= 1.10.0)

**Imports** Matrix

**Suggests** haven, testthat (>= 3.0.0)

**Config/testthat/edition** 3

**URL** <https://github.com/kylebutts/didimputation>

**License** MIT + file LICENSE

**NeedsCompilation** no

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**Repository** CRAN

**Date/Publication** 2026-03-09 06:40:10 UTC

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df\_het *Simulated data with two treatment groups and heterogenous effects*

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### Description

Generated using the following call: `did2s::gen_data(pane1 = c(1990, 2020), g1 = 2000, g2 = 2010, g3 = 0, te1 = 2, te2 = 1, te3 = 0, te_m1 = 0.05, te_m2 = 0.15, te_m3 = 0)`

### Usage

df\_het

### Format

A data frame with 31000 rows and 15 variables:

**unit** individual in panel data

**year** time in panel data

**g** the year that treatment starts

**dep\_var** outcome variable

**treat** T/F variable for when treatment is on

**rel\_year** year relative to treatment start. Inf = never treated.

**rel\_year\_binned** year relative to treatment start, but  $\leq -6$  and  $\geq 6$  are binned.

**unit\_fe** Unit FE

**year\_fe** Year FE

**error** Random error component

**te** Static treatment effect = te

**te\_dynamic** Dynamic treatment effect = te\_m

**state** State that unit is in

**group** String name for group

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df\_hom *Simulated data with two treatment groups and homogenous effects*

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### Description

Generated using the following call: `did2s::gen_data(pane1 = c(1990, 2020), g1 = 2000, g2 = 2010, g3 = 0, te1 = 2, te2 = 2, te3 = 0, te_m1 = 0, te_m2 = 0, te_m3 = 0)`

### Usage

df\_hom

**Format**

A data frame with 31000 rows and 15 variables:

**unit** individual in panel data

**year** time in panel data

**g** the year that treatment starts

**dep\_var** outcome variable

**treat** T/F variable for when treatment is on

**rel\_year** year relative to treatment start. Inf = never treated.

**rel\_year\_binned** year relative to treatment start, but  $\leq -6$  and  $\geq 6$  are binned.

**unit\_fe** Unit FE

**year\_fe** Year FE

**error** Random error component

**te** Static treatment effect = te

**te\_dynamic** Dynamic treatment effect = te\_m

**group** String name for group

**state** State that unit is in

**weight** Weight from runif()

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did\_imputation

*Borusyak, Jaravel, and Spiess (2021) Estimator*

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**Description**

Treatment effect estimation and pre-trend testing in staggered adoption diff-in-diff designs with an imputation approach of Borusyak, Jaravel, and Spiess (2021)

**Usage**

```
did_imputation(
  data,
  yname,
  gname,
  tname,
  idname,
  first_stage = NULL,
  wname = NULL,
  wtr = NULL,
  horizon = NULL,
  pretrends = NULL,
  cluster_var = NULL
)
```

**Arguments**

<code>data</code>	A <code>data.frame</code>
<code>yname</code>	String. Variable name for outcome. Use <code>fixest c()</code> syntax for multiple lhs, e.g. <code>"c(y1, y2)"</code> .
<code>gname</code>	String. Variable name for unit-specific date of treatment (never-treated should be zero or NA).
<code>tname</code>	String. Variable name for calendar period.
<code>idname</code>	String. Variable name for unique unit id.
<code>first_stage</code>	Formula for $Y(0)$ . Formula following <code>fixest::feols</code> . Fixed effects specified after <code>" "</code> . If not specified, then just unit and time fixed effects will be used.
<code>wname</code>	String. Variable name for estimation weights of observations. This is used in estimating $Y(0)$ and also augments treatment effect weights.
<code>wtr</code>	Character vector of treatment weight names (see <code>horizon</code> for standard static and event-study weights)
<code>horizon</code>	Integer vector of <code>event_time</code> or <code>TRUE</code> . This only applies if <code>wtr</code> is left as <code>NULL</code> . If specified, weighted averages/sums of treatment effects will be reported for each of these horizons separately (i.e. <code>tau0</code> for the treatment period, <code>tau1</code> for one period after treatment, etc.). If <code>TRUE</code> , all horizons are used. If <code>wtr</code> and <code>horizon</code> are null, then the static treatment effect is calculated.
<code>pretrends</code>	Integer vector or <code>TRUE</code> . Which pretrends to estimate. If <code>TRUE</code> , all pretrends are used.
<code>cluster_var</code>	String. Variable name for clustering groups. If not supplied, then <code>idname</code> is used as default.

**Details**

The imputation-based estimator is a method of calculating treatment effects in a difference-in-differences framework. The method estimates a model for  $Y(0)$  using untreated/not-yet-treated observations and predicts  $Y(0)$  for the treated observations  $\hat{Y}_{it}(0)$ . The difference between treated and predicted untreated outcomes  $Y_{it}(1) - \hat{Y}_{it}(0)$  serves as an estimate for the treatment effect for unit  $i$  in period  $t$ . These are then averaged to form average treatment effects for groups of  $(i, t)$ .

**Value**

A `data.frame` containing treatment effect term, estimate, standard error and confidence interval. This is in tidy format.

**Examples**

Load example dataset which has two treatment groups and homogeneous treatment effects

```
# Load Example Dataset
data("df_hom", package="didimputation")
```

**Static TWFE:**

You can run a static TWFE fixed effect model for a simple treatment indicator

```
did_imputation(data = df_hom, yname = "dep_var", gname = "g",
               tname = "year", idname = "unit")
#>      term estimate  std.error conf.low conf.high
#>   <char>    <num>      <num>   <num>   <num>
#> 1:  treat  2.024639  0.03243596  1.961065  2.088214
```

**Event Study:**

Or you can use relative-treatment indicators to estimate an event study estimate

```
did_imputation(data = df_hom, yname = "dep_var", gname = "g",
               tname = "year", idname = "unit", horizon=TRUE)
#>      term estimate  std.error conf.low conf.high
#>   <char>    <num>      <num>   <num>   <num>
#> 1:      0  2.117232  0.07368419  1.972811  2.261653
#> 2:      1  1.856536  0.07672104  1.706163  2.006909
#> 3:      2  1.986357  0.07137180  1.846468  2.126246
#> 4:      3  2.004843  0.07653409  1.854836  2.154850
#> 5:      4  1.950228  0.07543636  1.802372  2.098083
#> 6:      5  2.038302  0.07580288  1.889728  2.186875
#> 7:      6  2.031571  0.07223098  1.889999  2.173144
#> 8:      7  2.025286  0.07541719  1.877468  2.173104
#> 9:      8  1.976081  0.07493409  1.829210  2.122951
#> 10:     9  2.121434  0.07268404  1.978974  2.263895
#> 11:    10  2.087984  0.08271442  1.925864  2.250105
#> 12:    11  1.942825  0.11421421  1.718965  2.166685
#> 13:    12  1.940532  0.11200348  1.721005  2.160059
#> 14:    13  1.964569  0.11361969  1.741875  2.187264
#> 15:    14  2.023456  0.11753255  1.793092  2.253820
#> 16:    15  2.235051  0.12110086  1.997693  2.472409
#> 17:    16  2.178438  0.11552325  1.952013  2.404864
#> 18:    17  1.935576  0.11278311  1.714521  2.156631
#> 19:    18  2.134953  0.10993120  1.919488  2.350418
#> 20:    19  2.111984  0.11146282  1.893517  2.330451
#> 21:    20  1.925168  0.11214206  1.705370  2.144967
#>      term estimate  std.error conf.low conf.high
#>   <char>    <num>      <num>   <num>   <num>
```

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