

# simADHDsmart

## Description

This function simulates synthetic data to mirror desired characteristics of the Pelham ADHD SMART study. Created specifically for CATIE 2023.

## Usage

```
simADHDsmart(
  N = 150,
  baseline.params = list(p.odd = 0.4, m.severity = 0, p.priormed = 0.3, p.race = 0.8),
  Y0.coef = c(2, odd = -0.2, severity = -0.3),
  U.params = c(mu = 0, sd = 1),
  R.coef = c(-0.4, A1 = -0.1, `priormed:A1` = -0.2, U = 0.2),
  adherence.coef = c(-0.1, `priormed:A1` = -0.2, U = 0.1),
  event_time.coef = NULL,
  Y1.coef = c(2.5, A1 = -0.3, U = 0.1),
  Y2.baseline = c(3, odd = -0.3, severity = -0.4, priormed = 0, race = 0.5),
  Y2.tx1 = c(A1 = 0.3, `priormed:A1` = -1.4),
  Y2.n1 = c(R.resid = 0.8, adherence.resid = 1, U = 0.4),
  Y2.tx2 = c(A2 = -0.3, `A1:A2` = 0.1, `adherence:A2` = 1.2),
  sigma = 1
)
```

## Arguments

- |                 |   |
|-----------------|---|
| N               | number of observations to generate  |
| baseline.params | A list specifying the probability odd, mean severity, prob. priormed, prob. race  |
| Y0.coef         | A named vector specifying the linear model coefficients for baseline school performance. Can be a function of any past variables. See <i>Named Vectors</i>                        |
| U.params        | A vector specifying the normal distribution parameters $\mu$ and $\sigma$   |
| R.coef          | A named vector specifying the linear <b>probit model</b> coefficients for the probability of being a responder to first-stage treatment. Can be a function of any past variables. |
| adherence.coef  | A named vector specifying the linear <b>probit model</b> coefficients for the probability of being adherent to first-stage treatment. Can be a function of any past variables.    |

`event_time.coef` not implemented

`Y1.coef`

A named vector of linear model coefficients specifying the first-stage treatment causal effect on  $Y_1$  school performance

`Y2.baseline`

A named vector of linear model coefficients specifying the baseline covariate associations on end-of-study  $Y_2$  school performance

`Y2.tx1`

A named vector of linear model coefficients specifying the first-stage treatment causal effect on end-of-study  $Y_2$  school performance. Can be a function of any baseline moderators.

`Y2.n1`

A named vector of linear model coefficients specifying the nuisance associations on end-of-study  $Y_2$  school performance. Can be a function of any prior moderators. NOTE: must specify `R.resid` and `adherence.resid` which are the direct associations in the SNMM.

`Y2.tx2`

A named vector of linear model coefficients specifying the second-stage treatment causal effect, among non-responders, on end-of-study  $Y_2$  school performance. Can be a function of any baseline moderators. NOTE: all moderators are grand mean centered

`sigma`

gaussian noise added to  $Y_{0,1,2}$

## Value

A list with components

`data`

data.frame of observed variables

`DTRmean`

marginal embedded DTR means

## Named Vectors

The named vectors of coefficients use formula notation to specify interactions i.e. "`prior.med:A1`", which must be quoted. The order does not matter but spelling does. Note: if the first element is unnamed it is treated as the intercept term, else if all elements are named the intercept is omitted.

## Operating Characteristics

All baseline and moderator variable are grand mean centered. The nuisance terms `R.resid` and `adherence.resid` are residualized using the true probabilities.

### Default Structural Nested Mean Model:

School performance after first-stage

$$Y_1(a_1, a_2) = 2.5 - 0.3a_1 + 0.4U + \epsilon$$

School performance after second-stage

$$\begin{aligned} Y_2(a_1, a_2) = & 3 - 0.2odd - 0.3severity + 0 * priormed + .5race + \\ & (0.3 - 1.4priormed)a_1 + \\ & 0.8R.resid(a_1, u) + 1adherence.resid(a_1, u) \\ & + (1 - R)(-0.3 + 0.1a_1 + 1.2adherence)a_2 + 0.4U + \epsilon \end{aligned}$$

### Baseline Covariates:

- `odd`: binary, centered
- `severity`: standard normal
- `priormed`: binary, centered
- `race`: binary, centered
- `u`: an unknown, common cause of `R`, `adherence`, `Y1`, and `Y2`. Induces collider bias if naively conditioning on the time-varying covariates.

### First-Stage:

- Marginally, `BMOD(1)` is initially worse compared to `MED(-1)` after the first-stage  $Y_1$ , but better in the long run  $Y_2$
- IF on `priormed`, starting with `MED` is better

### Nuisance Associations

- `R`: positively associated with outcome
- `adherence`: positively associated with outcome
- `event-time`: or time to non-response has no effect

### Second-Stage (among non-responders only):

- Marginally, `AUG(-1)` is better compared to `INT(1)`

- Positive interaction for AUG if given MED, and INT if given BMOD
- IF adherent to first-stage, much better to INT; if non-adherent to first-stage, much better to AUG

## Internal functions

- `linearMult`: takes a named vector of coefficients and a `data.frame` and creates the necessary design matrix to multiply and return the function values.
- `sampleProbitMean`: generates observations from the posterior in order to empirically evaluate probability the integral.
- `getMarginalMeans`: takes the specified coefficients and returns the marginal means of the four embedded adaptive interventions, averaging over response and adherence.

## Author(s)

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