## Package 'noncomplyR'

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Type Package

Title Bayesian Analysis of Randomized Experiments with Non-Compliance

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Author Scott Coggeshall [aut, cre]

Maintainer Scott Coggeshall <sscogges@uw.edu>

Description Functions for Bayesian analysis of data from randomized experiments with noncompliance. The functions are based on the models described in Imbens and Rubin (1997) <doi:10.1214/aos/1034276631>. Currently only two types of outcome models are supported: binary outcomes and normally distributed outcomes. Models can be fit with and without the exclusion restriction and/or the strong access monotonicity assumption. Models are fit using the data augmentation algorithm as described in Tanner and Wong (1987) <doi:10.2307/2289457>.

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LazyData TRUE RoxygenNote 5.0.1 Imports MCMCpack (>= 1.4.0), stats Suggests knitr VignetteBuilder knitr NeedsCompilation no Repository CRAN

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cace

#### Description

cace takes a sample from the posterior distribution of the model parameters and computes the corresponding posterior distribution of the Complier Average Causal Effect.

#### Usage

cace(chain, outcome\_model, strong\_access)

#### Arguments

chain	a matrix containing the draws from the posterior distribution of the model pa- rameters. The matrix should either be the result of a call to compliance_chain or have the same structure as one.
outcome_model	a character string indicating which outcome model was used in fitting the model, either "binary" for a dichotomous outcome or "normal" for the Normal model.
strong_access	a logical indicating whether the strong access monotonicity assumption was made when fitting the model

#### Value

a vector containing the draws from the posterior distribution of the CACE

#### Examples

```
# CACE based on a subset of the vitaminA dataset
set.seed(4923)
chain <- compliance_chain(vitaminA[sample(1:nrow(vitaminA), 1000),], outcome_model = "binary",
exclusion_restriction = TRUE, strong_access = TRUE, n_iter = 10, n_burn = 1)
cace(chain, outcome_model = "binary", strong_access = TRUE)
```

```
# matrix representing the samples from the posterior distribution of the model parameters
posterior_mat <- matrix(rnorm(10*8, mean = 10), nrow = 10, ncol = 8)
cace(posterior_mat, "normal", strong_access = TRUE)
```

compliance\_chain Data Augmentation for Non-compliance analysis

#### Description

compliance\_chain fits a Bayesian non-compliance model by running a single chain of the data augmentation algorithm

#### Usage

```
compliance_chain(dat, outcome_model = NULL, exclusion_restriction = T,
  strong_access = T, starting_values = NULL, hyper_parameters = NULL,
  n_iter = 10000, n_burn = 1000)
```

#### Arguments

dat	a data frame. The first column of the data frame should be the outcome variable, the second column should be the treatment assignment variable, and the third column should be the indicator for the treatment actually received.	
outcome_model	a character string indicating how the outcome should be modeled. Either "nor- mal" for the normal model or "binary" for the binary model.	
exclusion_restriction		
	a logical value indicating whether the exclusion restriction assumption should be made.	
strong_access	a logical value indicating whether the strong access monotonicity assumption should be made.	
starting_values		
	the initial parameter values. If NULL, then the initial parameter values are based on either a random draw from the prior distribution (for the binary model) or sample statistics (for the normal model).	
hyper_parameters		
	a numerical vector containing the values that determine the prior distribution for the model parameters. If NULL, then the hyper parameters are chosen to give non-informative or reference priors.	
n_iter	number of iterations of the data augmentation algorithm to perform.	
n_burn	number of initial iterations to discard.	

#### Value

a matrix containing the draws from the posterior distribution.

#### Examples

```
# runs 10 iterations of the data augmentation algorithm on a subset of the vitaminA data
set.seed(4923)
```

```
compliance_chain(vitaminA[sample(1:nrow(vitaminA), 1000),], outcome_model = "binary",
exclusion_restriction = TRUE, strong_access = TRUE, n_iter = 10, n_burn = 1)
```

noncomplyR noncomplyR

#### Description

noncomplyR

#### Examples

## See function help files for examples

summarize\_chain Posterior Inference based on a Sample from the Posterior

#### Description

summarize\_chain provides posterior summaries based off a sample from the posterior distribution.

#### Usage

```
summarize_chain(chain, digits = 3)
```

#### Arguments

chain	a numeric vector containing the samples from the posterior distribution
digits	the number of decimal places

#### Value

a list containing the posterior mean, median, and quantile-based credible intervals calculated from the values in the chain.

#### Examples

```
# Suppose the posterior distribution was Normal(15, 5)
posterior_chain <- rnorm(100, 15, 5); summarize_chain(posterior_chain)</pre>
```

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vitaminA

#### Description

A dataset containing the results of an RCT investigating the effect of Vitamin A supplementation on mortality. The variables are as follows:

#### Usage

data(vitaminA)

#### Format

A data frame with 23682 rows and 3 variables

#### Details

- survived. indicator of survival (survived = 1) or death (survived = 0)
- vitaminA\_assigned. indicator of assignment to VitaminA (vitaminA\_assigned = 1) or placebo (vitaminA\_assigned = 0)
- vitaminA\_received. indicator of whether Vitamin A supplements were received (vitaminA\_received = 1) or not received (vitaminA\_received = 0)

#### Examples

## Not run: vitaminA

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