

# Package: hdbm (via r-universe)

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**Title** High Dimensional Bayesian Mediation Analysis

**Version** 0.9.0

**Description** Perform mediation analysis in the presence of high-dimensional mediators based on the potential outcome framework. High dimensional Bayesian mediation (HDBM), developed by Song et al (2018) <doi:10.1101/467399>, relies on two Bayesian sparse linear mixed models to simultaneously analyze a relatively large number of mediators for a continuous exposure and outcome assuming a small number of mediators are truly active. This sparsity assumption also allows the extension of univariate mediator analysis by casting the identification of active mediators as a variable selection problem and applying Bayesian methods with continuous shrinkage priors on the effects.

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 6.1.1

**LinkingTo** Rcpp, RcppArmadillo

**Imports** Rcpp

**Suggests** knitr, rmarkdown

**VignetteBuilder** knitr

**NeedsCompilation** yes

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**Repository** CRAN

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hdbm

*High Dimensional Bayesian Mediation***Description**

hdbm is a Bayesian inference method that uses continuous shrinkage priors for high-dimensional mediation analysis, developed by Song et al (2018). hdbm provides estimates for the regression coefficients as well as the posterior inclusion probability for ranking mediators.

**Usage**

```
hdbm(Y, A, M, C1, C2, beta.m, alpha.a, burnin, ndraws)
```

**Arguments**

Y	numeric outcome vector.
A	numeric exposure vector.
M	numeric matrix of mediators of Y and A.
C1	numeric matrix of extra covariates in the outcome model
C2	numeric matrix of extra covariates in the mediator model
beta.m	numeric vector of initial beta.m in the outcome model
alpha.a	numeric vector of initial alpha.a in the mediator model
burnin	number of iterations to run the MCMC before sampling
ndraws	number of draws to take from MCMC after the burnin period

**Details**

hdbm uses two regression models for the two conditional relationships,  $Y|A, M, C1$  and  $M|A, C2$ . For the outcome model, hdbm uses

$$Y = M\beta_M + A * \beta_A + C1 * \beta_C Y + \epsilon_Y$$

For the mediator model, hdbm uses the model

$$M = A * \alpha_A + C2 * \alpha_C 2 + \epsilon_M$$

For high dimensional tractability, hdbm employs continuous Bayesian shrinkage priors to select mediators and makes the two following assumptions: First, it assumes that all the potential mediators contribute small effects in mediating the exposure-outcome relationship. Second, it assumes that only a small proportion of mediators exhibit large effects ("active" mediators). hdbm uses a Metropolis-Hastings within Gibbs MCMC to generate posterior samples from the model.

**Value**

hdbm returns a list with 11 elements (each of length `ndraws`), sampled from the burned in MCMC:

**beta.m** Outcome model mediator coefficients

**r1** Whether or not each `beta.m` belongs to the larger normal component (1) or smaller normal component (0)

**alpha.a** Mediator model exposure coefficients

**r3** Whether or not each `alpha.a` belongs to the larger normal component (1) or smaller normal component (0)

**beta.a** `beta.a` coefficient

**pi.m** Proportion of non zero `beta.m` coefficients

**pi.a** Proportion of non zero `alpha.a` coefficients

**sigma.m0** standard deviation of the smaller normal component for mediator-outcome coefficients (`beta.m`)

**sigma.m1** standard deviation of the larger normal component for mediator-outcome coefficients (`beta.m`)

**sigma.ma0** Standard deviation of the smaller normal component for exposure-mediator coefficients (`alpha.a`)

**sigma.ma1** Standard deviation of the larger normal component for exposure-mediator coefficients (`alpha.a`)

**Author(s)**

Alexander Rix

**References**

Yanyi Song, Xiang Zhou et al. Bayesian Shrinkage Estimation of High Dimensional Causal Mediation Effects in Omics Studies. bioRxiv [10.1101/467399](https://doi.org/10.1101/467399)

**Examples**

```
library(hdbm)

Y <- hdbm.data$y
A <- hdbm.data$a

# grab the mediators from the example data.frame
M <- as.matrix(hdbm.data[, paste0("m", 1:100)], nrow(hdbm.data))

# We just include the intercept term in this example.
C <- matrix(1, 1000, 1)
beta.m <- rep(0, 100)
alpha.a <- rep(0, 100)

set.seed(12345)
hdbm.out <- hdbm(Y, A, M, C, C, beta.m, alpha.a,
```

```
burnin = 1000, ndraws = 100)

# Which mediators are active?
active <- which(colSums(hdbm.out$r1 * hdbm.out$r3) > 50)
colnames(M)[active]
```

---

hdbm.data

*Synthetic example data for hdbm*

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### **Description**

Synthetic example data for hdbm

### **Usage**

hdbm.data

### **Format**

A data.frame with 1000 observations on 102 variables:

**y** Numeric response variable.

**a** Numeric exposure variable.

**m[1-100 ]** Numeric mediator variables

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