## Package 'BayesCOOP'

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| <ul> <li>Title Implements the BayesCOOP Methodology for Supervised Multimodal Integration</li> <li>Version 0.1.0</li> <li>Description Combines jittered group spike-and-slab LASSO regularization with intermediate fusion to enable integrative learning across multiple data modalities. For uncertainty quantification, BayesCOOP applies the Bayesian bootstrap to generate approximate posterior samples by performing maximum a posteriori (MAP) estimation on jittered and resampled multimodal datasets. Currently, only continuous outcomes are supported.</li> </ul> |  |          |                                      |  |
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#### **Description**

Type Package

This function implements the BayesCOOP methodology for supervised multimodal integration. It combines jittered group spike-and-slab LASSO regularization with intermediate fusion to enable integrative learning across multiple data modalities. For uncertainty quantification, BayesCOOP applies the Bayesian bootstrap to generate approximate posterior samples by performing maximum a posteriori (MAP) estimation on jittered and resampled multimodal datasets. Currently, only continuous outcomes are supported.

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### Usage

```
bayesCoop(
  data_train,
  data_test,
  family = "gaussian",
  ss = c(0.05, 1),
  group = TRUE,
  bb = TRUE,
  alpha_dirich = 1,
  bbiters = 1100,
  bburn = 100,
  maxit = 100,
  filter = TRUE,
  abd_thresh = 0,
  prev_thresh = 0.1,
  Warning = TRUE,
  verbose = TRUE,
  control = list()
)
```

## **Arguments**

control

when bb = FALSE.

| data_train   | a list of feature_table, sample_metadata and feature_metadata from training data (unstandardized). See <i>IntegratedLearner</i> for more details. |
|--------------|---|
| data_test    | a list of feature_table, sample_metadata and feature_metadata from testing data (unstandardized). See <i>IntegratedLearner</i> for more details.  |
| family       | currently supports only Gaussian family. Default: "gaussian".   |
| SS           | a length-2 numeric vector giving the spike/sLab scales $c(s0,s1)$ with $s0 < s1$ . Default: $c(0.05,1)$ .   |
| group        | logical. If TRUE, predictors are grouped by modality and given a group spike-and-slab prior. If FALSE, no grouping is used. Default: TRUE.        |
| bb           | logical. If TRUE, run full Bayesian bootstrap inference; if FALSE, run MAP estimation over a user-supplied rho grid in control. Default: TRUE.    |
| alpha_dirich | Dirichlet concentration for Bayesian bootstrap weights. Default: 1.   |
| bbiters      | number of Bayesian bootstrap iterations. Default: 1100.   |
| bbburn       | number of burn-in iterations discarded from the bootstrap. Default: 100.  |
| maxit        | maximum EM iterations in the inner optimizer. Default: 100.   |
| filter       | logical. If TRUE, apply abundance/prevalence filtering to features. Default: TRUE.  |
| abd_thresh   | minimum abundance threshold for keeping a feature. Default: 0.  |
| prev_thresh  | minimum prevalence threshold (proportion of samples above abd_thresh). Default: 0.1.  |
| Warning      | logical. If TRUE, emit non-convergence warnings. Default: TRUE.   |
| verbose      | logical. If TRUE, print iteration counts and runtime. Default: TRUE.  |
|              |   |

a named list with element rho, giving one or more candidate rho values to try

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#### Value

If bb = TRUE, a list with:

y\_samples posterior predictive draws (approximate)

y\_pred posterior predictive median for each held-out sample

mspe mean squared prediction error on test data beta\_samples posterior draws of regression coefficients

beta\_postmed posterior median coefficients

rho\_samples posterior draws of consensus penalty rho

rho\_postmed posterior median of rho

errVar\_samples posterior draws of residual variance

time runtime in minutes

If bb = FALSE, a list with:

y\_pred predicted response values on test data
mspe mean squared prediction error on test data
beta\_MAP MAP estimate of regression coefficients

rho\_MAP selected rho (minimizing MSPE over control\$rho)

time runtime in minutes

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