Package 'iBART'

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Title Iterative Bayesian Additive Regression Trees Descriptor Selection Method

Version 1.0.0

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Description A statistical method based on Bayesian Additive Regression Trees with Global Standard Error Permutation Test (BART-G.SE) for descriptor selection and symbolic regression. It finds the symbolic formula of the regression function y=f(x) as described in Ye, Senftle, and Li (2023) <arXiv:2110.10195>.

URL https://github.com/mattsheng/iBART

BugReports https://github.com/mattsheng/iBART/issues License GPL (>= 3) Encoding UTF-8 LazyData true VignetteBuilder knitr RoxygenNote 7.2.3 Depends R (>= 4.0.0) Imports bartMachine (>= 1.2.6), glmnet (>= 4.1-1), foreach, stats Suggests knitr, rmarkdown, ggplot2, ggpubr SystemRequirements Java (>= 8.0) NeedsCompilation no Author Shengbin Ye [aut, cre, cph] (<https://orcid.org/0000-0001-8767-2595>), Meng Li [aut] Repository CRAN Date/Publication 2023-11-14 17:40:02 UTC

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catalysis

Single-Atom Catalysis Data

Description

Single-Atom Catalysis Data

Usage

catalysis

Format

A list with 4 objects:

X Primary feature matrix: physical properties of transition metals and oxide supports

y Reponse variable: binding energy of metal/oxide pairs

head Column names of X

unit Unit of columns of X

generate_unit A helper function to generate unit for iBART input

Description

A helper function to generate unit for iBART input

Usage

```
generate_unit(unit, dimension)
```

Arguments

unit	A vector of unit of the primary features. For example, unit <- c("cm", "s"). Then
	the unit of $x1$ is centimeter and the unit of $x2$ is second.
dimension	A vector of dimension of the units. For example, unit <- c("cm", "s") and di-
	mension $\langle c(2, 1) \rangle$ mean that the unit of x_1 is square centimeter and the unit of
	x2 is second.

Value

A list that contains unit and dimension information.

iBART

Description

Finds a symbolic formula for the regression function y = f(X) using (y, X) as inputs.

Usage

```
iBART(
 X = NULL,
 y = NULL,
 head = NULL,
 unit = NULL,
 BART_var_sel_method = "global_se",
 num_trees = 20,
  num_burn_in = 10000,
  num_iterations_after_burn_in = 5000,
  num_reps_for_avg = 10,
  num_permute_samples = 50,
  type.measure = "deviance",
  nfolds = 10,
  nlambda = 100,
  relax = FALSE,
  gamma = c(0, 0.25, 0.5, 0.75, 1),
  opt = c("binary", "unary", "binary"),
  sin_cos = FALSE,
  apply_pos_opt_on_neg_x = TRUE,
  hold = 0,
 pre_screen = TRUE,
  corr_screen = TRUE,
  out_sample = FALSE,
  train_idx = NULL,
  train_ratio = 1,
 Lzero = TRUE,
  parallel = FALSE,
 K = ifelse(Lzero, 5, 0),
 aic = FALSE,
  standardize = TRUE,
 writeLog = FALSE,
 verbose = TRUE,
  count = NULL,
  seed = NULL
```

)

Arguments

Х

Input matrix of primary features X.

У	Response variable y.
head	Optional: name of primary features.
unit	Optional: units and their respective dimensions of primary features. This is used to perform dimension analysis for generated descriptors to avoid generating unphylical descriptors, such as $size + size^2$. See generate_dimension() for details.
BART_var_sel_m	ethod
	Variable selection criterion used in BART. Three options are available: (1) "global_se", (2) "global_max", (3) "local". The default is "global_se". See var_selection_by_permute in R package bartMachine for more detail.
num_trees	BART parameter: number of trees to be grown in the sum-of-trees model. If you want different values for each iteration of BART, input a vector of length equal to number of iterations. Default is num_trees = 20.
num_burn_in	BART parameter: number of MCMC samples to be discarded as "burn-in". If you want different values for each iteration of BART, input a vector of length equal to number of iterations. Default is num_burn_in = 10000.
num_iterations	_after_burn_in
	BART parameter: number of MCMC samples to draw from the posterior distribution of $hatf(x)$. If you want different values for each iteration of BART, input a vector of length equal to number of iterations. Default is num_iterations_after_burn_in = 5000.
num_reps_for_a	
	BART parameter: number of replicates to over over to for the BART model's variable inclusion proportions. If you want different values for each iteration of BART, input a vector of length equal to number of iterations. Default is num_reps_for_avg = 10.
num_permute_sa	mples
	BART parameter: number of permutations of the response to be made to gen- erate the "null" permutation distribution. If you want different values for each iteration of BART, input a vector of length equal to number of iterations. Default is num_permute_samples = 50.
type.measure	glmnet parameter: loss to use for cross-validation. The default is type.measure="deviance", which uses squared-error for Gaussian models (a.k.a type.measure="mse" there). type.measure="mae" (mean absolute error) can be used also.
nfolds	glmnet parameter: number of folds - default is 10. Smallest value allowable is nfolds=3.
nlambda	glmnet parameter: the number of lambda values - default is 100.
relax	glmnet parameter: If TRUE, then CV is done with respect to the mixing param- eter gamma as well as lambda. Default is relax=FALSE.
gamma	glmnet parameter: the values of the parameter for mixing the relaxed fit with the regularized fit, between 0 and 1; default is gamma = $c(0, 0.25, 0.5, 0.75, 1)$
opt	A vector of operation order. For example, opt = c("unary", "binary", "unary") will apply unary operators, then binary operators, then unary operators. Available operator sets are "unary", "binary", and "all", where "all" is the union of "unary" and "binary".

iBART

sin_cos	Logical flag for using $sin(\pi * x)$ and $cos(\pi * x)$ to generate descriptors. This is useful if you think there is periodic relationship between predictors and response. Default is sin_cos = FALSE.
apply_pos_opt_	on_neg_x
	Logical flag for applying non-negative-valued operators, such as \sqrt{x} and $log(x)$, when some values of x is negative. If apply_pos_opt_on_neg_x == TRUE, apply absolute value operator first then non-negative-valued operator, i.e. generate $\sqrt{ x }$ and $log(x)$ instead. Default is apply_pos_opt_on_neg_x = TRUE.
hold	Number of iterations to hold. This allows iBART to run consecutive operator transformations before screening. Note hold = 0 is equivalent to no skipping of variable selection in each iBART iterations. It should be less than iter.
pre_screen	Logical flag for pre-screening the primary features X using BART. Only selected primary features will be used to generate descriptors. Note that pre_screen = FALSE is equivalent to hold = 1.
corr_screen	Logical flag for screening out primary features that are independet of the response variable y .
out_sample	Logical flag for out of sample assessment. Default is out_sample = FALSE.
train_idx	Numerical vector storing the row indices for training data. Please set out_sample = TRUE if you supplied train_idx.
train_ratio	Proportion of data used to train model. Value must be between (0,1]. This is only needed when out_sample = TRUE and train_idx == NULL. Default is train_ratio = 1.
Lzero	Logical flag for L-zero variable selection. Default is Lzero = TRUE.
parallel	Logical flag for parallel L-zero variable selection. Default is parallel = FALSE.
К	If Lzero == TRUE, K sets the maximum number of descriptors to be selected.
aic	If Lzero == TRUE, logical flag for selecting best number of descriptors using AIC. Possible number of descriptors are $1 \le k \le K$.
standardize	Logical flag for data standardization prior to model fitting in BART and LASSO. Default is standardize = TRUE.
writeLog	Logical flag for writing log file to working directory. The log file will contain in- formation such as the descriptors selected by iBART, RMSE of the linear model build on the selected descriptors, etc. Default is writeLog = FALSE.
verbose	Logical flag for printing progress to console. Default is verbose = TRUE.
count	Internal parameter. Default is count = NULL.
seed	Optional: sets the seed in both R and Java. Default is seed = NULL which does not set the seed in R nor Java.

Value

A list of iBART output.

iBART_model	The LASSO output of the last iteration of iBART. The predictors with non-zero coefficient are called the iBART selected descriptors.
X_selected	The numerical values of the iBART selected descriptors.

descr	descriptor_names					
		The names of the iBART selected descriptors.				
coeff	ficients	Coefficients of the iBART model. The first element is an intercept.				
X_tra	ain	The training matrix used in the last iteration.				
X_tes	st	The testing matrix used in the last iteration.				
iBAR1	ſ_gen_size	The number of descriptors generated by iBART in each iteration.				
iBAR1	[_sel_size	The number of descriptors selected by iBART in each iteration.				
iBAR1	[_in_sample	e_RMSE				
		In sample RMSE of the LASSO model.				
iBAR1	[_out_samp]	le_RMSE				
		Out of sample RMSE of the LASSO model if out_sample == TRUE.				
Lzero	_models	The l_0 -penalized regression models fitted on the iBART selected descriptors for $1 \le k \le K$.				
Lzero	o_names	The name of the best kD descriptors selected by the l_0 -penalized regression model for $1 \le k \le K$.				
Lzero	o_in_sample	e_RMSE				
		In sample RMSE of the l_0 -penalized regression model for $1 \le k \le K$.				
Lzero	o_out_samp	le_RMSE				
		Out of sample RMSE of the l_0 -penalized regression model for $1 \le k \le K$ if out_sample == TRUE.				
Lzero	_AIC_mode	1				
		The best l_0 -penalized regression model selected by AIC.				
Lzero	o_AIC_names	S				
		The best kD descriptors where $1 \le k \le K$ is chosen via AIC.				
Lzero	o_AIC_in_sa					
		In sample RMSE of the best l_0 -penalized regression models chosen by AIC.				
Lzero	o_AIC_out_s	sample_RMSE				
		Out of sample RMSE of the best l_0 -penalized regression models chosen by AIC if out_sample == TRUE.				
runti	ime	Runtime in second.				

Author(s)

Shengbin Ye

References

Ye, S., Senftle, T.P., and Li, M. (2023) *Operator-induced structural variable selection for identifying materials genes*, https://arxiv.org/abs/2110.10195.

Description

iBART result in the real data vignette

Usage

iBART_real_data

Format

A list of iBART outputs

iBART_model A cv.glmnet object storing the iBART selected model ...

iBART_sim *iBART Simulation Result*

Description

iBART result in the simulation vignette

Usage

iBART_sim

Format

A list of iBART outputs

iBART_model A cv.glmnet object storing the iBART selected model ...

k_var_model

Description

Best subset selection for linear regression

Usage

```
k_var_model(
 X_train,
 y_train,
 X_test = NULL,
 y_test = NULL,
 k = 1,
 parallel = FALSE
)
```

Arguments

X_train	The design matrix used during training.
y_train	The response variable used during training.
X_test	The design matrix used during testing. Default is X_test = NULL and full data will be used to train the best subset linear regression model.
y_test	The response variable used during testing. Default is $y_{test} = NULL$ and full data will be used to train the best subset linear regression model.
k	The maximum number of predictors allowed in the model. For example, $k = 5$ will produce the best model 5 predictors.
parallel	Logical flag for parallelization. Default is parallel = FALSE.

Value

A list of outputs.

models	An 1m object storing the best k-predictor linear model.
names	The variable name of the best k predictors.
rmse_in	In-sample RMSE of the model.
rmse_out	Out-of-sample RMSE of the model.

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