

Package: LSAMBA (via r-universe)

May 20, 2026

Type Package

Title Lasso-SAMBA Algorithm

Version 2024.1.1

Maintainer Auriane Gabaut <aurianegabaut@gmail.com>

Description Provides a lasso-based method for building mechanistic models using the SAMBA algorithm (Stochastic Approximation for Model Building Algorithm) (M Prague, M Lavielle (2022) <[doi:10.1002/psp4.12742](https://doi.org/10.1002/psp4.12742)>). The package extends the 'Rsmlx' package (version 2024.1.0) to better handle high-dimensional data. It relies on the 'Monolix' software (version 2024R1; see <<https://monolixsuite.slp-software.com/monolix/2024R1/>>), which must be installed beforehand.

SystemRequirements 'Monolix'
(<<https://monolixsuite.slp-software.com/monolix/2024R1/>>)

License GPL (>= 3)

Encoding UTF-8

LazyData true

RoxygenNote 7.3.2

Imports doParallel, dplyr, ggplot2, grDevices, latex2exp, MASS, parallel, sharp, stringr

Depends R (>= 2.10), Rsmlx (>= 2024.1.0), foreach

Roxygen list(markdown = TRUE)

Config/pak/sysreqs cmake libglpk-dev libicu-dev libxml2-dev

Repository <https://aurianegbt.r-universe.dev>

Date/Publication 2026-01-19 15:43:02 UTC

RemoteUrl <https://github.com/aurianegbt/lsamba>

RemoteRef HEAD

RemoteSha 748e14045749712cf44c3422b7cda102fbeb8894

Contents

buildmlx	2
getMLXdir	5

Index	6
--------------	----------

buildmlx	<i>buildmlx: Automatic statistical model building</i>
----------	---

Description

buildmlx uses SAMBA (Stochastic Approximation for Model Building Algorithm), an iterative procedure to accelerate and optimize the process of model building by identifying at each step how best to improve some of the model components. This method allows to find the optimal statistical model which minimizes some information criterion in very few steps.

Usage

```
buildmlx(
  project = NULL,
  final.project = NULL,
  model = "all",
  prior = NULL,
  weight = NULL,
  coef.w1 = 0.5,
  paramToUse = "all",
  covToTest = "all",
  covToTransform = "none",
  center.covariate = FALSE,
  criterion = "BICc",
  linearization = FALSE,
  ll = TRUE,
  test = FALSE,
  direction = NULL,
  steps = 1000,
  n.full = 10,
  max.iter = 20,
  explor.iter = 2,
  fError.min = 0.001,
  seq.cov = FALSE,
  seq.cov.iter = 0,
  seq.corr = TRUE,
  p.max = if (buildMethod == "stepAIC") {
    0.1
  } else {
    1
  },
),
```

```

p.min = c(0.075, 0.05, 0.1),
print = TRUE,
nb.model = 1,
nfolds = 5,
alpha = 1,
nSS = 1000,
buildMethod = "lasso",
FDR_thr = 0.1
)

```

Arguments

project	a string: the initial Monolix project
final.project	a string: the final Monolix project (default adds "_built" to the original project)
model	components of the model to optimize c("residualError", "covariate", "correlation"), (default="all")
prior	list of prior probabilities for each component of the model (default=NULL)
weight	list of penalty weights for each component of the model (default=NULL)
coef.w1	multiplicative weight coefficient used for the first iteration only (default=0.5)
paramToUse	list of parameters possibly function of covariates (default="all")
covToTest	components of the covariate model that can be modified (default="all")
covToTransform	list of (continuous) covariates to be log-transformed (default="none")
center.covariate	TRUE/FALSE center the covariates of the final model (default=FALSE)
criterion	penalization criterion to optimize c("AIC", "BIC", "BICc", gamma) (default=BICc)
linearization	TRUE/FALSE whether the computation of the likelihood is based on a linearization of the model (default=FALSE)
ll	TRUE/FALSE compute the observe likelihood and the criterion to optimize at each iteration
test	TRUE/FALSE perform additional statistical tests for building the model (default=FALSE)
direction	for stepAIC method, method for covariate search c("full", "both", "backward", "forward"), (default="full" or "both")
steps	for stepAIC method, maximum number of iteration for stepAIC (default=1000)
n.full	for stepAIC method, maximum number of covariates for an exhaustive comparison of all possible covariate models (default=10)
max.iter	maximum number of iterations (default=20)
explor.iter	number of iterations during the exploratory phase (default=2)
fError.min	minimum fraction of residual variance for combined error model (default = 1e-3)
seq.cov	TRUE/FALSE whether the covariate model is built before the correlation model
seq.cov.iter	number of iterations before building the correlation model (only when seq.cov=F, default=0)

seq.corr	TRUE/FALSE whether the correlation model is built iteratively (default=TRUE)
p.max	maximum p-value used for removing non significant relationships between covariates and individual parameters (default=0.1 for stepAIC and 1 for lasso)
p.min	vector of 3 minimum p-values used for testing the components of a new model (default=c(0.075, 0.05, 0.1))
print	TRUE/FALSE display the results (default=TRUE)
nb.model	number of models to display at each iteration (default=1)
nolds	for lasso method, number of folds (default=10)
alpha	for lasso method, the elasticnet mixing parameter, between 0 and 1. alpha=1 is the lasso penalty, alpha=0 is the ridge penalty.
nSS	for lasso method, number of resampling iterations for stability selection.
buildMethod	the method used to build the covariate model (default="lasso")
FDR_thr	for lasso method, upper-bounds in FDP of calibrated stability selection (default=0.1)

Details

For covariates model building, covariate selection can be achieved by stepAIC, as the original SAMBA algorithm was implemented in RsmIx package (Prague and Lavielle, 2020 ; Mihaljevic, 2023) and by a lasso approach enhanced by stability selection (Bodinier et al., 2023).

Value

a new Monolix project with a new statistical model.

References

- Prague M, Lavielle M. SAMBA: A novel method for fast automatic model building in nonlinear mixed-effects models. *CPT Pharmacometrics Syst Pharmacol*. 2022; 11: 161-172. doi:10.1002/psp4.12742
- Bodinier B, Filippi S, Haugdahl Nøst T, Chiquet J, Chadeau-Hyam M. Automated calibration for stability selection in penalised regression and graphical models. *Journal of the Royal Statistical Society Series C: Applied Statistics*. 2023 ; 72: 1375–1393. doi:10.1093/jrssc/qlad058
- Mihaljevic F (2023). RsmIx: R Speaks 'Monolix'. R package version 2023.1.5, <https://CRAN.R-project.org/package=RsmIx>.

Examples

```
## Not run:
project <- getMLXdir()

res = buildmlx(project = project,
               buildMethod = "lasso",
               model='covariate',
               test=FALSE)

getIndividualParameterModel()

## End(Not run)
```

getMLXdir	<i>Get monolix demo project path.</i>
-----------	---------------------------------------

Description

Get monolix demo project path.

Usage

```
getMLXdir()
```

Value

path to the monolix demo from 'Monolix' software

Examples

```
## Not run:  
print(getMLXdir())  
  
## End(Not run)
```

Index

`buildmlx`, [2](#)

`getMLXdir`, [5](#)