Package 'BASiCStan'

December 2, 2025

Title Stan implementation of BASiCS

Version 1.13.0

Date 2024-02-14

Description Provides an interface to infer the parameters of BASiCS using the variational inference (ADVI), Markov chain Monte Carlo (NUTS), and maximum a posteriori (BFGS) inference engines in the Stan programming language. BASiCS is a Bayesian hierarchical model that uses an adaptive Metropolis within Gibbs sampling scheme. Alternative inference methods provided by Stan may be preferable in some situations, for example for particularly large data or posterior distributions with difficult geometries.

License GPL-3

Encoding UTF-8

VignetteBuilder knitr

Roxygen list(markdown = TRUE)

RoxygenNote 7.2.1

Biarch true

Depends R (>= 4.2), BASiCS, rstan (>= 2.18.1)

Imports methods, glmGamPoi, scran, scuttle, stats, utils, SingleCellExperiment, SummarizedExperiment, Rcpp (>= 0.12.0), RcppParallel (>= 5.0.1), rstantools (>= 2.1.1)

LinkingTo BH (>= 1.66.0), Rcpp (>= 0.12.0), RcppEigen (>= 0.3.3.3.0), RcppParallel (>= 5.0.1), rstan (>= 2.18.1), StanHeaders (>= 2.18.0)

SystemRequirements GNU make

biocViews ImmunoOncology, Normalization, Sequencing, RNASeq, Software, GeneExpression, Transcriptomics, SingleCell, DifferentialExpression, Bayesian, CellBiology

URL https://github.com/Alanocallaghan/BASiCStan

BugReports https://github.com/Alanocallaghan/BASiCStan/issues

Suggests testthat (>= 3.0.0), knitr, rmarkdown

Config/testthat/edition 3

git_url https://git.bioconductor.org/packages/BASiCStan

git_branch devel

git_last_commit a442827

git_last_commit_date 2025-10-29

Repository Bioconductor 3.23

Date/Publication 2025-12-02

Author Alan O'Callaghan [aut, cre],
Catalina Vallejos [aut]

Maintainer Alan O'Callaghan <alan.ocallaghan@outlook.com>

Contents

BASiCStan-package 2
BASiCStan 3
Stan2BASiCS 4

Index **6**

`BASiCStan-package` *The 'BASiCStan' package.*

Description

Provides an interface to infer the parameters of BASiCS using the variational inference (ADVI), Markov chain Monte Carlo (NUTS), and maximum a posteriori (BFGS) inference engines in the Stan programming language. BASiCS is a Bayesian hierarchical model that uses an adaptive Metropolis within Gibbs sampling scheme. Alternative inference methods provided by Stan may be preferable in some situations, for example for particularly large data or posterior distributions with difficult geometries. See also BASiCS_MCMC

References

Stan Development Team (2020). RStan: the R interface to Stan. R package version 2.21.2. https://mc-stan.org

Vallejos, Marioni and Richardson (2015). PLoS Computational Biology. https://doi.org/10.1371/journal.pcbi.1004333.

Vallejos, Richardson and Marioni (2016). Genome Biology. https://doi.org/10.1186/s13059-016-0930-3

Eling et al (2018). Cell Systems. https://doi.org/10.1016/j.cels.2018.06.011

BASiCStan *Stan implementation of BASiCS.*

Description

The stan programming language enables the use of MAP, VB, and HMC inference. Only the regression mode featuring a joint prior between mean and overdispersion parameters is implemented

Usage

```
BASiCStan(
  Data,
  Method = c("vb", "sampling", "optimizing"),
  WithSpikes = length(altExpNames(Data)) > 0,
  Regression = TRUE,
  BatchInfo = Data$BatchInfo,
  L = 12,
  PriorMu = c("EmpiricalBayes", "uninformative"),
  NormFactorFun = scran::calculateSumFactors,
  ReturnBASiCS = TRUE,
  Verbose = TRUE,
  ...
)
```

Arguments

Data	SingleCellExperiment object
Method	Inference method. One of: "vb" for Variational Bayes, "sampling" for Hamiltonian Monte Carlo, "optimizing" for or maximum *a posteriori* estimation.
WithSpikes	Do the data contain spike-in genes? See BASiCS for details. When WithSpikes=FALSE, the cell-specific scaling normalisation factors are fixed; use NormFactorFun to specify how size factors should be generated or extracted.
Regression	Use joint prior for mean and overdispersion parameters? Included for compatibility with BASiCS_MCMC, but only TRUE is supported.
BatchInfo	Vector describing which batch each cell is from.
L	Number of regression terms (including slope and intercept) to use in joint prior for mu and delta.
PriorMu	Type of prior to use for mean expression. Default is "EmpiricalBayes", but "uninformative" is the prior used in Eling et al. and previous work.
NormFactorFun	Function that returns cell-specific scaling normalisation factors. See computeSumFactors for details on the default.
ReturnBASiCS	Should the object be converted into a BASiCS_Chain object?
Verbose	Should output of the stan commands be printed to the terminal?
...	Passed to vb or sampling.

Value

An object of class BASiCS_Chain.

Examples

```
library("BASiCS")
sce <- BASiCS_MockSCE(NGenes = 10, NCells = 10)

fit_spikes <- BASiCStan(sce, tol_rel_obj = 1)

## uses fixed scaling normalisation factors
fit_nospikes <- BASiCStan(sce, WithSpikes = FALSE, tol_rel_obj = 1)
```

Stan2BASiCS *Convert Stan fits to BASiCS_Chain objects.*

Description

Convert Stan fits to BASiCS_Chain objects.

Usage

```
Stan2BASiCS(
  x,
  gene_names = attr(x, "gene_names"),
  cell_names = attr(x, "cell_names"),
  size_factors = attr(x, "size_factors")
)
```

Arguments

x	A stan object
gene_names, cell_names	Gene and cell names. The reason this argument exists is that by default, stan fit parameters are not named. NOTE: this must be the same order as the data supplied to BASiCStan.
size_factors	Cell-specific scaling normalisation factors, to be stored as part of the chain object when WithSpikes=FALSE.

Value

A BASiCS_Chain object.

Examples

```
library("BASiCS")
sce <- BASiCS_MockSCE(NGenes = 10, NCells = 10)

fit_spikes <- BASiCStan(sce, ReturnBASiCS = FALSE, tol_rel_obj = 1)
summary(fit_spikes)
```

Index

BASiCS_Chain, *3*, *4*
BASiCS_MCMC, *2*, *3*
BASiCStan, 3, *4*
BASiCStan-package, 2

computeSumFactors, *3*

Stan2BASiCS, 4