

# Package ‘MIMOSA’

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**Type** Package

**Title** Mixture Models for Single-Cell Assays

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**Description** Modeling count data using Dirichlet-multinomial and beta-binomial mixtures with applications to single-cell assays.

**License** Artistic-2.0

**VignetteBuilder** knitr

**Imports** methods, Formula, data.table, pracma, MCMCpack, coda, modeest, testthat, Rcpp, scales, Kmisc

**Suggests** parallel, knitr

**Depends** R (>= 3.0.2), MASS, plyr, reshape, Biobase, ggplot2

**LinkingTo** Rcpp, RcppArmadillo

**LazyLoad** yes

**LazyData** yes

**biocViews** FlowCytometry, CellBasedAssays

**NeedsCompilation** yes

## R topics documented:

MIMOSA-package . . . . .	2
.fitMCMC . . . . .	2
asinh_trans . . . . .	3
BetaMixResult-class . . . . .	4
boxplotMIMOSAResultList . . . . .	4
combine.MIMOSA . . . . .	5
ConstructMIMOSAEExpressionSet . . . . .	6
countsTable . . . . .	7

fdr . . . . .	8
getZ . . . . .	9
ICS . . . . .	10
MDMix . . . . .	11
MIMOSA . . . . .	11
MIMOSAExpressionSet . . . . .	12
MIMOSAResult . . . . .	13
MIMOSAResult-class . . . . .	13
pData,MIMOSAResult-method . . . . .	13
print.MIMOSAResultList . . . . .	14
show . . . . .	15
volcanoPlot . . . . .	15

<b>Index</b>	<b>17</b>
--------------	-----------

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MIMOSA-package

*MIMOSA: Mixture Models for Single Cell Assays*

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## Description

MIMOSA implements mixtures of Dirichlet-multinomial or Beta-binomial models for paired count data from single-cell assays that typically arise in immunological studies. It can be used for ICS (Intracellular Cytokine Staining) assays to detect vaccine responders, for example, or to detect changes in proportions of cells expressing a gene, such as in Fluidigm Biomark Single-cell gene expression.

## References

Greg Finak, Andrew McDavid, Pratip Chattopadhyay, Maria Dominguez, Stephen C De Rosa, Mario Roederer, Raphael Gottardo Mixture Models for Single Cell Assays with Applications to Vaccine Studies Biostatistics, 2013, <http://biostatistics.oxfordjournals.org/content/early/2013/07/24/biostatistics.kxt024.abstract>

## See Also

[MIMOSA](#), [ConstructMIMOSAExpressionSet](#)

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.fitMCMC

*Fit the MIMOSA model via MCMC*

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## Description

This is an internal function that fits the MIMOSA model via MCMC. It is called from MIMOSA

**Usage**

```
.fitMCMC(data, inits = NULL, iter = 250000, burn = 50000, thin = 1,
tune = 100, outfile = basename(tempfile(tmpdir = ".", fileext = ".dat")),
alternative = "greater", UPPER = 0.5, LOWER = 0.15, FAST = TRUE,
EXPRATE = 1e-04, pXi = c(1, 1), seed = 10)
```

**Arguments**

<b>data</b>	a list with elements names 'n.stim' and 'n.unstim', the stimulated and unstimulated counts. Must be at least of dimension 2.
<b>inits</b>	the initialization parameters for the MCMC routine. Can be initialized from MDMix with initonly=TRUE.
<b>iter</b>	the number of Mote Carlo iterations
<b>burn</b>	the number of burn-in iterations
<b>thin</b>	The thinning interval
<b>tune</b>	the number of iterations used for tuning the step size
<b>outfile</b>	the output file name
<b>alternative</b>	either 'greater' or 'not equal' for fitting the one-sided or two-sided MIMOSA model, respectively.
<b>UPPER</b>	tuning parameter for the upper bound on the acceptance ratio of each parameter
<b>LOWER</b>	tuning parameter for the lower bound on the acceptance ratio of each parameter
<b>FAST</b>	TRUE, FALSE. Use the heuristic (FAST=TRUE) for fitting a one-sided model rather than recomputing the normalization constant via MCMC for each step.
<b>EXPRATE</b>	the mean of the prior distribution for the model hyperparameters.
<b>pXi</b>	is the prior on the w, beta(1,1) by default).
<b>seed</b>	numeric random seed

asinh\_trans

asinh\_trans

**Description**

Arcsinh transform for ggplot2

**Usage**

asinh\_trans(c)

**Arguments**

<b>c</b>	numeric cofactor for asinh trasnform. Default 1.
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**Details**

Arcsinh transform for use with coord\_trans in ggplot2

**Value**

transform

**Author(s)**

Greg Finak

**BetaMixResult-class**     *The output of fitting Beta-Binomial EM implementation BetaMix.*

**Description**

BetaMix will return an object of this class.

**boxplotMIMOSAResultList**  
*boxplotMIMOSAResultList*

**Description**

Boxplots of MIMOSA

**Usage**

```
boxplotMIMOSAResultList(data, title = "A Boxplot", x_axis_category = NULL,
                        cofactor = 5000, line = TRUE, threshold = 0.005)
```

**Arguments**

data	MIMOSAResultList
title	character Title of the plot.
x_axis_category	name the column of the phenoData frame for the x-axis of the boxplots.
cofactor	integer cofactor of the arcsinhTransform for the y axis.
line	logical whether or not to connect points from the same subject
threshold	numeric the FDR threshold (q-value) at which to classify responders as a separate category.

**Details**

Generate boxplots for MIMOSA positivity calls.

**Value**

ggplot object.

**Author(s)**

Greg Finak

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combine.MIMOSA

*Combine MIMOSAResultList objects*

---

**Description**

Combine two or more MIMOSAResultList objects

**Usage**

combine.MIMOSA(x, y, ...)

**Arguments**

x	MIMOSAResultList
y	MIMOSAResultList
...	additional MIMOSAResultList objects

**Details**

Combines two or more MIMOSAResultList objects. The method is light on error checking so the results should be from the same MIMOSAEExpressionSet object.

**Value**

a MIMOSAResultList

**Author(s)**

Greg Finak

**ConstructMIMOSAExpressionSet***A wrapper for constructing an Expression Set for MIMOSA***Description**

Calls a series of other functions that will reshape and refactor the data frame into the right format for use by MIMOSA Standardized for use with internal SCHARP data sets. We provide some default arguments as examples. Currently slow, and very much prototype code.

**Usage**

```
ConstructMIMOSAExpressionSet(thisdata, reference = quote(STAGE %in% "CTRL" &
PROTEIN %in% "Media+cells"), measure.columns = c("Neg", "Pos"),
other.annotations = setdiff(colnames(thisdata), measure.columns),
default.cast.formula = component ~ ..., .variables = quote(.(PTID, TESTDT,
ASSAYID, PLATEID)), featureCols = 1, ref.append.replace = "_NEG")
```

**Arguments**

<code>thisdata</code>	is the input data frame
<code>reference</code>	is an expression that evaluates to a logical vector which specifies the observations in the data frame that are to be used for the negative control or reference set
<code>measure.columns</code>	is a character vector that specifies which columns hold the observed counts
<code>other.annotations</code>	is a character vector that specifies which additional columns in the data frame should be included in the returned data. By default we take everything, but you could specify only relevant phenotypic information.
<code>default.cast.formula</code>	is a formula that tells reshape how to recast the data frame so that rows correspond to different measured components and columns correspond to samples. By default <code>component~...</code> will put the components as the rows (i.e. positive and negative cell counts) and all measured phenotypic information on the columns.
<code>.variables</code>	is a dotted list that specifies the variable names (columns of the data frame) by which to group the data when organizing stimulated and unstimulated observations. i.e. PTID x ANTIGEN x TCELLSUBSET x TESTDT, or something else for your own data.
<code>featureCols</code>	is a numeric vector that specifies the indices of the columns to be used to name the features. If the casting formula is <code>component~...</code> then there is only one feature column (and it is the first one), so <code>featureCols = 1</code> , by default.
<code>ref.append.replace</code>	the terminating character string in the column names of the negative controls. It will be replaced with <code>_REF</code> for 'reference'

## Examples

```
data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')
```

countsTable

*Extract the table of counts from a MIMOSA model*

## Description

Extract the table of counts from a MIMOSA model

## Usage

```
countsTable(object, proportion = FALSE)

## S4 method for signature 'MIMOSAResult'
countsTable(object, proportion = FALSE)

## S4 method for signature 'MCMCResult'
countsTable(object, proportion = FALSE)

## S4 method for signature 'MDMixResult'
countsTable(object, proportion = FALSE)

## S3 method for class 'MIMOSAResultList'
countsTable(object, proportion = FALSE)

## S4 method for signature 'MIMOSAResultList'
countsTable(object, proportion = FALSE)
```

## Arguments

object	a MIMOSAResult
proportion	logical return the counts or the proportions

## Value

a data.frame of counts to which the model was fit.  
 a data.frame of counts for the stimulated and unstimulated samples

## Examples

```

data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')

result<-MIMOSA(NSUB+CYTNUM~UID+TCELLSUBSET+CYTOKINE|ANTIGEN,
  data=E, method='EM',
  subset=RefTreat%in%'Treatment'&ANTIGEN%in%'ENV',
  ref=ANTIGEN%in%'ENV'&RefTreat%in%'Reference')
head(countsTable(result))
head(countsTable(result,proportion=TRUE))

```

fdr

*Compute the fdr (q-value) from posterior probabilities*

## Description

Given the z's from a MIMOSA model, calculates the q-values for each observation.

## Usage

```

fdr(z)

## S3 method for class 'matrix'
fdr(z)

## S3 method for class 'MIMOSAResult'
fdr(z)

## S3 method for class 'MIMOSAResultList'
fdr(z)

```

## Arguments

**z** matrix of posterior probabilities, or a MIMOSAResult, or MIMOSAResultList

## Value

a vector of q-values or a list of vectors of q-values.

## Examples

```
data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')
result<-MIMOSA(NSUB+CYTNUM~UID+TCELLSUBSET+CYTOKINE|ANTIGEN,
  data=E, method='EM',
  subset=RefTreat%in%'Treatment'&ANTIGEN%in%'ENV',
  ref=ANTIGEN%in%'ENV'&RefTreat%in%'Reference')
qvalues<-fdr(result)
```

getZ

*Extract the posterior probabilities of response from a MIMOSA model*

## Description

Extract the posterior probabilities of response from a MIMOSA model

Extract the component weights from a MIMOSA model

## Usage

```
getZ(x)

## S3 method for class 'MIMOSAResultList'
getZ(x)

## S3 method for class 'MIMOSAResult'
getZ(x)

getW(x)

## S3 method for class 'MIMOSAResultList'
getW(x)

## S3 method for class 'MIMOSAResult'
getW(x)
```

## Arguments

x                   output from a MIMOSA model

## Value

a matrix of posterior probabilities  
 a vector of component weights

## Examples

```

data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')

result<-MIMOSA(NSUB+CYTNUM~UID+TCELLSUBSET+CYTOKINE|ANTIGEN,
  data=E, method='EM',
  subset=RefTreat%in%'Treatment'&ANTIGEN%in%'ENV',
  ref=ANTIGEN%in%'ENV'&RefTreat%in%'Reference')
getZ(result)
data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')

result<-MIMOSA(NSUB+CYTNUM~UID+TCELLSUBSET+CYTOKINE|ANTIGEN,
  data=E, method='EM',
  subset=RefTreat%in%'Treatment'&ANTIGEN%in%'ENV',
  ref=ANTIGEN%in%'ENV'&RefTreat%in%'Reference')
getW(result)

```

ICS

*Stimulated and unstimulated T-cell counts for an ICS assay*

## Description

A data set containing T-cell counts for various stimulations and cytokines in an ICS assay.

## Format

A data frame with 3960 rows

## Details

- pos. The positive cell counts
- neg. The negative cell counts
- fname. The feature name (cytokine) measured
- parent. The parent T-cell population
- antigen. The antigen stimulation for this sample
- ID. The subject ID

---

MDMix*EM fitting of the Multinomial Dirichlet MIMOSA model.*

---

**Description**

This function fits the multinomial dirichelt MIMOSA model using EM. It can also be used to initialize the model parameters for the MCMC model.

**Usage**

```
MDMix(data = NULL, modelmatrix = NULL, alternative = "greater",
      initonly = FALSE)
```

**Arguments**

data	The observed data
modelmatrix	a model matrix specifying which components should be computed
alternative	either 'greater' or 'not equal' to fit the one-sided or two-sided model.
initonly	TRUE or FALSE to return just the initialization parameters.

**Value**

An object of class `MDMixResult`

**Author(s)**

Greg Finak TODO filtering of pu>ps needs to be corrected here.

---

MIMOSA*Fit a MIMOSA Model*

---

**Description**

This method fits a MIMOSA model to count data stored in an ExpressionSet object.

**Usage**

```
MIMOSA(formula, data, ...)
```

**Arguments**

formula	describing the features on the lhs and the phenodata on the rhs, supporting extended formula interface with conditioning.
data	an ExpressionSet object with features on rows and samples (labelled with phenoData) on columns.
...	additional arguments

## Details

The ExpressionSet should be fully annotated with featureData and phenoData. For ICS data, for example, features would be positive and negative counts for different cytokine producing cell subsets (i.e. IFNg\_pos, IFNg\_neg) The formula lhs should contain features and the rhs should contain phenotypic variable. See the vignette for an example.

## Value

an object of type MIMOSAResult

## See Also

[MIMOSA-package](#) [ConstructMIMOSAExpressionSet](#) [MIMOSAResult](#)

## Examples

```
data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='\_REF')

result<-MIMOSA(NSUB+CYTNUM~UID+TCELLSUBSET+CYTOKINE|ANTIGEN,
  data=E, method='EM',
  subset=RefTreat%in%'Treatment'&ANTIGEN%in%'ENV',
  ref=ANTIGEN%in%'ENV'&RefTreat%in%'Reference')
```

MIMOSAExpressionSet     *Construct an ExpressionSet for MIMOSA*

## Description

Starting from a reshaped data frame in the correct format, construct an ExpressionSet object that can be used with MIMOSA.

## Usage

```
MIMOSAExpressionSet(df, featureCols)
```

## Arguments

- |             |  |
|-------------|--|
| df          | a data.frame that is in the correct form           |
| featureCols | the indices of the columns that identify features. |

## Details

The featureCols will be used to construct feature names, and these columns will be dropped from the exprs matrix. The column names are assumed to have names that contain '\_' characters separating phenotypic characteristics. These would be generated automatically if the data frame was constructed with 'reshape'. They are used to construct the phenoData for the expression set

## Examples

```
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')
```

MIMOSAResult

*Stores the result of a MIMOSA fitted model*

## Description

MIMOSA returns an object of MIMOSAResult irrespective of which method / implementation is used to fit the data.

MIMOSAResult-class

*Stores the result of a MIMOSA fitted model*

## Description

MIMOSA returns an object of MIMOSAResult irrespective of which method / implementation is used to fit the data.

pData,MIMOSAResult-method

*pData extract the phenoData table from a MIMOSA result*

## Description

pData extract the phenoData table from a MIMOSA result

**Usage**

```
## S4 method for signature 'MIMOSAResult'
pData(object)

## S4 method for signature 'MDMixResult'
pData(object)

## S4 method for signature 'MCMCResult'
pData(object)

pData.MIMOSAResultList(object)

## S4 method for signature 'MIMOSAResultList'
pData(object)
```

**Arguments**

**object**            is the MIMOSAResult returned from a call to MIMOSA

**Details**

Extracts the phenoData data.frame from a MIMOSAResult object

**Value**

an object of type `data.frame`

**print.MIMOSAResultList**  
*Print a MIMOSAResultList*

**Description**

Print a summary of the list of results returned by a call to MIMOSA

**Usage**

```
## S3 method for class 'MIMOSAResultList'
print(x, ...)
```

**Arguments**

<b>x</b>	a MIMOSAResultList
<b>...</b>	additional arguments passed down

---

show

*show*

---

### Description

Show a MIMOSAResultList

### Usage

```
## S4 method for signature 'MIMOSAResult'  
show(object)
```

### Arguments

object            MIMOSAResultList

### Details

Show a summary of a MIMOSAResultList.

---

volcanoPlot

*Volcano plot for a MIMOSA model*

---

### Description

Plots effect size vs posterior probability of response from a MIMOSAResultList, faceting by the conditioning variables.

### Usage

```
volcanoPlot(x, effect_expression = NA, facet_var = NA, threshold = 0.01)
```

### Arguments

x                A MIMOSAResultList  
effect\_expression

an expression that defines the effect size. Usually a function of the stimulated and unstimulated proportions from countsTable(x, proportion=TRUE)

facet\_var        an expression defining the faceting in ggplot parlance. i.e. ~ faceting + variables

threshold        a numeric value between [0,1] for coloring significant observations (based on the q-value)

### See Also

[countsTable](#)

**Examples**

```
data(ICS)
E<-ConstructMIMOSAExpressionSet(ICS,
  reference=ANTIGEN%in%'negctrl',measure.columns=c('CYTNUM','NSUB'),
  other.annotations=c('CYTOKINE','TCELLSUBSET','ANTIGEN','UID'),
  default.cast.formula=component~UID+ANTIGEN+CYTOKINE+TCELLSUBSET,
  .variables=.(TCELLSUBSET,CYTOKINE,UID),
  featureCols=1,ref.append.replace='_REF')

result<-MIMOSA(NSUB+CYTNUM~UID+TCELLSUBSET+CYTOKINE|ANTIGEN,
  data=E, method='EM',
  subset=RefTreat%in%'Treatment'&ANTIGEN%in%'ENV',
  ref=ANTIGEN%in%'ENV'&RefTreat%in%'Reference')
volcanoPlot(result,CYTNUM-CYTNUM_REF)
```

# Index

\*Topic **datasets**  
    ICS, 10  
    .fitMCMC, 2  
  
    asinh\_trans, 3  
  
BetaMixResult-class, 4  
boxplotMIMOSAResultList, 4  
  
combine.MIMOSA, 5  
ConstructMIMOSAExpressionSet, 2, 6, 12  
countsTable, 7, 15  
countsTable, MCMCRResult-method  
    (countsTable), 7  
countsTable, MDMixResult-method  
    (countsTable), 7  
countsTable, MIMOSAResult-method  
    (countsTable), 7  
countsTable, MIMOSAResultList-method  
    (countsTable), 7  
countsTable.MIMOSAResultList  
    (countsTable), 7  
  
fdr, 8  
  
getW(getZ), 9  
getZ, 9  
  
ICS, 10  
  
MDMix, 11  
MIMOSA, 2, 11  
MIMOSA, formula, ExpressionSet-method  
    (MIMOSA), 11  
MIMOSA-package, 2  
MIMOSAExpressionSet, 12  
MIMOSAResult, 12, 13  
MIMOSAResult-class, 13  
  
pData, MCMCRResult-method  
    (pData, MIMOSAResult-method), 13  
pData, MIMOSAResult-method, 13  
pData, MIMOSAResultList-method  
    (pData, MIMOSAResult-method), 13  
pData.MIMOSAResultList  
    (pData, MIMOSAResult-method), 13  
print.MIMOSAResultList, 14  
  
show, 15  
show, MIMOSAResult-method (show), 15  
  
volcanoPlot, 15