

# Package ‘CytoGLMM’

November 6, 2024

**Type** Package

**Title** Conditional Differential Analysis for Flow and Mass Cytometry Experiments

**Version** 1.14.0

**Description** The CytoGLMM R package implements two multiple regression strategies: A bootstrapped generalized linear model (GLM) and a generalized linear mixed model (GLMM). Most current data analysis tools compare expressions across many computationally discovered cell types. CytoGLMM focuses on just one cell type. Our narrower field of application allows us to define a more specific statistical model with easier to control statistical guarantees. As a result, CytoGLMM finds differential proteins in flow and mass cytometry data while reducing biases arising from marker correlations and safeguarding against false discoveries induced by patient heterogeneity.

**License** LGPL-3

**URL** <https://christofseiler.github.io/CytoGLMM>,  
<https://github.com/ChristofSeiler/CytoGLMM>

**BugReports** <https://github.com/ChristofSeiler/CytoGLMM/issues>

**Encoding** UTF-8

**LazyData** true

**Imports** stats, methods, BiocParallel, RColorBrewer, cowplot,  
doParallel, dplyr, factoextra, flexmix, ggplot2, magrittr,  
mbest, pheatmap, stringr, strucchange, tibble, ggrepel, MASS,  
logging, Matrix, tidyr, caret, rlang, grDevices

**Suggests** knitr, rmarkdown, testthat, BiocStyle

**VignetteBuilder** knitr

**RoxygenNote** 7.2.3

**biocViews** FlowCytometry, Proteomics, SingleCell, CellBasedAssays,  
CellBiology, ImmunoOncology, Regression, StatisticalMethod,  
Software

**git\_url** <https://git.bioconductor.org/packages/CytoGLMM>

**git\_branch** RELEASE\_3\_20

**git\_last\_commit** 019171c

**git\_last\_commit\_date** 2024-10-29

**Repository** Bioconductor 3.20

**Date/Publication** 2024-11-05

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cytoflexmix	<i>Logistic mixture regression</i>
-------------	------------------------------------

---

## Description

Logistic mixture regression

## Usage

```
cytoflexmix(
  df_samples_subset,
  protein_names,
  condition,
  group = "donor",
  cell_n_min = Inf,
  cell_n_subsample = 0,
  ks = seq_len(10),
  num_cores = 1
)
```

**Arguments**

df_samples_subset	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
condition	The column name of the condition variable
group	The column name of the group variable
cell_n_min	Remove samples that are below this cell counts threshold
cell_n_subsample	Subsample samples to have this maximum cell count
ks	A vector of cluster sizes
num_cores	Number of computing cores

**Value**

A list of class `cytoglm` containing

flexmixfits	list of <code>flexmix</code> objects
df_samples_subset	possibly subsampled <code>df_samples_subset</code> table
protein_names	input protein names
condition	input condition variable
group	input group names
cell_n_min	input <code>cell_n_min</code>
cell_n_subsample	input <code>cell_n_subsample</code>
ks	input <code>ks</code>
num_cores	input <code>num_cores</code>

**Examples**

```

set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
mix_fit <- CytoGLMM::cytoflexmix(df,
                                protein_names = protein_names,
                                condition = "condition",
                                group = "donor",
                                ks = 2)

mix_fit

```

cytoglm

*Fit GLM with bootstrap resampling***Description**

Fit GLM with bootstrap resampling

**Usage**

```
cytoglm(
  df_samples_subset,
  protein_names,
  condition,
  group = "donor",
  covariate_names = NULL,
  cell_n_min = Inf,
  cell_n_subsample = 0,
  num_boot = 100,
  num_cores = 1
)
```

**Arguments**

df_samples_subset	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
condition	The column name of the condition variable
group	The column name of the group variable
covariate_names	The column names of covariates
cell_n_min	Remove samples that are below this cell counts threshold
cell_n_subsample	Subsample samples to have this maximum cell count
num_boot	Number of bootstrap samples
num_cores	Number of computing cores

**Value**A list of class `cytoglm` containing

tb_coef	coefficient table
df_samples_subset	possibly subsampled df_samples_subset table
protein_names	input protein names
condition	input condition variable
group	input group names
covariate_names	input covariates

```

cell_n_min      input cell_n_min
cell_n_subsample
                 input cell_n_subsample

unpaired        true if unpaired samples were provided as input
num_boot        input num_boot
num_cores       input num_cores
formula_str     formula use in the regression model

```

### Examples

```

set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
glm_fit <- CytoGLMM::cytoglm(df,
                             protein_names = protein_names,
                             condition = "condition",
                             group = "donor",
                             num_boot = 10) # in practice >=1000

glm_fit

```

---

cytoglmm

*Fit GLMM with method of moments*

---

### Description

Fit GLMM with method of moments

### Usage

```

cytoglmm(
  df_samples_subset,
  protein_names,
  condition,
  group = "donor",
  covariate_names = NULL,
  cell_n_min = Inf,
  cell_n_subsample = 0,
  num_cores = 1
)

```

### Arguments

```

df_samples_subset      Data frame or tibble with proteins counts, cell condition, and group information
protein_names          A vector of column names of protein to use in the analysis
condition              The column name of the condition variable
group                  The column name of the group variable
covariate_names        The column names of covariates

```

cell\_n\_min      Remove samples that are below this cell counts threshold  
 cell\_n\_subsample      Subsample samples to have this maximum cell count  
 num\_cores      Number of computing cores

### Value

A list of class `cytoglm` containing

`glmmfit`      `mbest` object  
`df_samples_subset`      possibly subsampled `df_samples_subset` table  
`protein_names`      input protein names  
`condition`      input condition variable  
`group`      input group names  
`covariate_names`      input covariates  
`cell_n_min`      input `cell_n_min`  
`cell_n_subsample`      input `cell_n_subsample`  
`num_cores`      input `num_cores`

### Examples

```

set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
glmm_fit <- CytoGLMM::cytoglmm(df,
                                protein_names = protein_names,
                                condition = "condition",
                                group = "donor")

glmm_fit

```

---

cytogroup

*Group-specific fixed effects model*

---

### Description

Group-specific fixed effects model

### Usage

```

cytogroup(
  df_samples_subset,
  protein_names,
  condition,
  group = "donor",
  cell_n_min = Inf,
  cell_n_subsample = 0
)

```

**Arguments**

df_samples_subset	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
condition	The column name of the condition variable
group	The column name of the group variable
cell_n_min	Remove samples that are below this cell counts threshold
cell_n_subsample	Subsample samples to have this maximum cell count

**Value**

A list of class `cytoglm` containing

groupfit	<code>glm</code> object
df_samples_subset	possibly subsampled <code>df_samples_subset</code> table
protein_names	input protein names
condition	input condition variable
group	input group names
cell_n_min	input <code>cell_n_min</code>
cell_n_subsample	input <code>cell_n_subsample</code>

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
group_fit <- CytoGLMM::cytogroup(df,
                                protein_names = protein_names,
                                condition = "condition",
                                group = "donor")

group_fit
```

---

 cytostab

---

*Evaluate parameter stability with respect to gating scheme*


---

**Description**

Evaluate parameter stability with respect to gating scheme

**Usage**

```

cytostab(
  df_samples_subset,
  protein_names,
  condition,
  group = "donor",
  cell_n_min = Inf,
  cell_n_subsample = 0
)

```

**Arguments**

df_samples_subset	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
condition	The column name of the condition variable
group	The column name of the group variable
cell_n_min	Remove samples that are below this cell counts threshold
cell_n_subsample	Subsample samples to have this maximum cell count

**Value**

A data frame

**Examples**

```

set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
stab <- CytoGLMM::cytostab(df,
                           protein_names = protein_names,
                           condition = "condition",
                           group = "donor")

stab

```

---

cyto\_check

*Check if input to cytoxxx function have errors*

---

**Description**

Check if input to cytoxxx function have errors

**Usage**

```
cyto_check(cell_n_subsample, cell_n_min, protein_names)
```



**Arguments**

cell\_n\_subsample      Subsample samples to have this maximum cell count  
cell\_n\_min            A vector of column names of protein to use in the analysis  
protein\_names        A vector of column names of protein to use in the analysis

**Value**

NULL.

---

generate_data	<i>Generate dataset for vignettes and simulation studies</i>
---------------	--

---

**Description**

Generate dataset for vignettes and simulation studies

**Usage**

```
generate_data()
```

**Value**

`tibble` data frame

**Examples**

```
set.seed(23)  
df <- generate_data()  
str(df)  
df
```

---

glmm_moment	<i>Generalized linear mixed model with maximum likelihood</i>
-------------	---

---

**Description**

Generalized linear mixed model with maximum likelihood

**Usage**

```
glmm_moment(  
  df_samples,  
  protein_names,  
  response,  
  group = "donor",  
  covariate_names = NULL,  
  num_cores = 1  
)
```

**Arguments**

df_samples	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
response	The column name of the condition variable
group	The column name of the group variable
covariate_names	The column names of covariates
num_cores	Number of computing cores

**Value**

`mbest` object

---

is_unpaired	<i>Check if samples match or paired on condition</i>
-------------	--

---

**Description**

Check if samples match or paired on condition

**Usage**

```
is_unpaired(df_samples_subset, condition, group)
```

**Arguments**

df_samples_subset	Data frame or tibble with proteins counts, cell condition, and group information
condition	The column name of the condition variable
group	The column name of the group variable

**Value**

A boolean

---

plot.cytoflexmix	<i>Plot all components of mixture regression</i>
------------------	--

---

**Description**

Plot all components of mixture regression

**Usage**

```
## S3 method for class 'cytoflexmix'
plot(x, k = NULL, separate = FALSE, ...)
```

**Arguments**

x	A cytoflexmix class
k	Number of clusters
separate	create two separate <code>ggplot2</code> objects
...	Other parameters

**Value**

`ggplot2` object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
mix_fit <- CytoGLMM::cytoflexmix(df,
                                protein_names = protein_names,
                                condition = "condition",
                                group = "donor",
                                ks = 2)

plot(mix_fit)
```

---

plot.cytoglm                      *Plot bootstraped coefficients*

---

**Description**

Plot bootstraped coefficients

**Usage**

```
## S3 method for class 'cytoglm'
plot(x, order = FALSE, separate = FALSE, ...)
```

**Arguments**

x	A cytoglm class
order	Order the markers according to the mangintute of the coefficients
separate	create two separate <code>ggplot2</code> objects
...	Other parameters

**Value**

`ggplot2` object



---

plot.cytogroup	<i>Plot fixed coefficients of group-specific fixed effects model</i>
----------------	--

---

**Description**

Plot fixed coefficients of group-specific fixed effects model

**Usage**

```
## S3 method for class 'cytgroup'
plot(x, order = FALSE, separate = FALSE, ...)
```

**Arguments**

x	A cytoglmm class
order	Order the markers according to the magnitude of the coefficients
separate	create two separate <code>ggplot2</code> objects
...	Other parameters

**Value**

`ggplot2` object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
group_fit <- CytoGLMM::cytgroup(df,
                                protein_names = protein_names,
                                condition = "condition",
                                group = "donor")

plot(group_fit)
```

---

plot_coeff	<i>Helper function to plot regression coefficient</i>
------------	---

---

**Description**

Helper function to plot regression coefficient

**Usage**

```
plot_coeff(
  tb,
  title_str,
  title_str_right,
  xlab_str,
  redline = 0,
  order = FALSE,
  separate = FALSE
)
```

**Arguments**

tb	A data frame
title_str	Title string for summary plot
title_str_right	Title for bootstrap sample plot
xlab_str	Label on x-axis
redline	Point on x-axis to draw the red line
order	Order the markers according to the magnitude of the coefficients
separate	Plot both summary and bootstrap samples

**Value**

[ggplot2](#) object or list of two objects if separate is true

---

plot_heatmap	<i>Heatmap of median marker expression</i>
--------------	--

---

**Description**

Heatmap of median marker expression

**Usage**

```
plot_heatmap(
  df_samples,
  sample_info_names,
  protein_names,
  arrange_by_1,
  arrange_by_2 = "",
  cluster_cols = FALSE,
  fun = median
)
```

**Arguments**

df_samples	Data frame or tibble with proteins counts, cell condition, and group information
sample_info_names	Column names that contain information about the cell, e.g. donor, condition, file name, or cell type
protein_names	A vector of column names of protein to use in the analysis
arrange_by_1	Column name
arrange_by_2	Column name
cluster_cols	Apply hierarchical cluster to columns
fun	Summary statistics of marker expression

**Value**

`pheatmap` object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
CytoGLMM::plot_heatmap(df,
                        protein_names = protein_names,
                        sample_info_names = c("donor", "condition"),
                        arrange_by_1 = "condition")
```

---

plot\_lda

*LDA on marker expression*

---

**Description**

LDA on marker expression

**Usage**

```
plot_lda(
  df_samples,
  protein_names,
  group,
  cor_scaling_factor = 1,
  arrow_color = "black",
  marker_color = "black",
  marker_size = 5
)
```

**Arguments**

df_samples	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
group	The column name of the group variable
cor_scaling_factor	Scaling factor of circle of correlations
arrow_color	Color of correlation circle
marker_color	Colors of marker names
marker_size	Size of marker names

**Value**

`ggplot2` object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
df$condition <- rep(c("A", "B", "C", "D"), each = length(df$condition)/4)
CytoGLMM::plot_lda(df,
                    protein_names = protein_names,
                    group = "condition",
                    cor_scaling_factor = 2)
```

---

plot\_mds

*MDS on median marker expression*

---

**Description**

MDS on median marker expression

**Usage**

```
plot_mds(
  df_samples,
  protein_names,
  sample_info_names,
  color,
  sample_label = ""
)
```

**Arguments**

df_samples	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
sample_info_names	Column names that contain information about the cell, e.g. donor, condition, file name, or cell type
color	Column name
sample_label	Column name



**Value**

cowplot object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
CytoGLMM::plot_mds(df,
                    protein_names = protein_names,
                    sample_info_names = c("donor", "condition"),
                    color = "condition")
```

---

plot\_model\_selection *Plot model selection to choose number optimal number of clusters*

---

**Description**

Plot model selection to choose number optimal number of clusters

**Usage**

```
plot_model_selection(fit, k = NULL)
```

**Arguments**

fit	A cytoflexmix class
k	Number of clusters

**Value**

cowplot object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
mix_fit <- CytoGLMM::cytoflexmix(df,
                                protein_names = protein_names,
                                condition = "condition",
                                group = "donor",
                                ks = 1:2)

plot_model_selection(mix_fit)
```

---

plot_prcomp	<i>Plot PCA of subsampled data using ggplot</i>
-------------	---

---

**Description**

Plot PCA of subsampled data using ggplot

**Usage**

```
plot_prcomp(
  df_samples,
  protein_names,
  color_var = "treatment",
  subsample_size = 10000,
  repel = TRUE
)
```

**Arguments**

df_samples	Data frame or tibble with proteins counts, cell condition, and group information
protein_names	A vector of column names of protein to use in the analysis
color_var	A column name
subsample_size	Subsample per color_var variable
repel	Repel labels

**Value**

cowplot object

**Examples**

```
set.seed(23)
df <- generate_data()
protein_names <- names(df)[3:12]
df <- dplyr::mutate_at(df, protein_names, function(x) asinh(x/5))
CytogLM::plot_prcomp(df,
  protein_names = protein_names,
  color_var = "condition")
```

---

print.cytoglm	<i>Extact and print bootstrap GLM fit</i>
---------------	---

---

**Description**

Extact and print bootstrap GLM fit

**Usage**

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



---

remove_samples	<i>Remove samples based on low cell counts</i>
----------------	--

---

**Description**

Remove samples based on low cell counts

**Usage**

```
remove_samples(df_samples_subset, condition, group, unpaired, cell_n_min)
```

**Arguments**

df_samples_subset	Data frame or tibble with proteins counts, cell condition, and group information
condition	The column name of the condition variable
group	The column name of the group variable
unpaired	true if unpaired samples were provided as input
cell_n_min	Remove samples that are below this cell counts threshold

**Value**

NULL.

---

summary.cytoglm	<i>Extact and calculate p-values of bootstrap GLM fit</i>
-----------------	---

---

**Description**

Extact and calculate p-values of bootstrap GLM fit

**Usage**

```
## S3 method for class 'cytoglm'
summary(object, method = "BH", ...)
```

**Arguments**

object	A cytoglm class
method	Multiple comparison adjustment method
...	Other parameters

**Value**

[tibble](#) data frame



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