

# Package ‘gINTomics’

April 3, 2025

**Title** Multi-Omics data integration

**Version** 1.3.0

**Description** gINTomics is an R package for Multi-Omics data integration and visualization.

gINTomics is designed to detect the association between the expression of a target and of its regulators, taking into account also their genomics modifications such as Copy Number Variations (CNV) and methylation.

What is more, gINTomics allows integration results visualization via a Shiny-based interactive app.

**License** AGPL-3

**biocViews** GeneExpression, RNASeq, Microarray, Visualization, CopyNumberVariation, GeneTarget

**Encoding** UTF-8

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.3.1

**Imports** BiocParallel, biomaRt, OmnipathR, edgeR, ggplot2, ggridges, gtools, MultiAssayExperiment, plyr, stringi, stringr, SummarizedExperiment, methods, stats, reshape2, randomForest, limma, org.Hs.eg.db, org.Mm.eg.db, BiocGenerics, GenomicFeatures, ReactomePA, clusterProfiler, dplyr, AnnotationDbi, TxDb.Hsapiens.UCSC.hg38.knownGene, TxDb.Mmusculus.UCSC.mm10.knownGene, shiny, GenomicRanges, ggtree, shinydashboard, plotly, DT, MASS, InteractiveComplexHeatmap, ComplexHeatmap, visNetwork, shiny.gosling, ggvenn, RColorBrewer, utils, grDevices, callr, circlize

**Depends** R (>= 4.4.0)

**LazyData** false

**Suggests** BiocStyle, knitr, rmarkdown, testthat (>= 3.0.0)

**Config/testthat/edition** 3

**VignetteBuilder** knitr

**BugReports** <https://github.com/angelovelle96/gINTomics/issues>

**URL** <https://github.com/angelovelle96/gINTomics>

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

**git\_branch** devel

**git\_last\_commit** 3ba788e

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

**Repository** Bioconductor 3.21

**Date/Publication** 2025-04-03

**Author** Angelo Velle [cre, aut] (ORCID:  
<https://orcid.org/0000-0002-4010-6390>),  
 Francesco Patane' [aut] (ORCID:  
<https://orcid.org/0009-0001-8619-447X>),  
 Chiara Romualdi [aut] (ORCID: <https://orcid.org/0000-0003-4792-9047>)

**Maintainer** Angelo Velle <angelo.velle@unipd.it>

## Contents

gINTomics-package . . . . .	3
create_multiassay . . . . .	3
dot_plotly . . . . .	4
extract_model_res . . . . .	5
mirna_hsa . . . . .	7
mmultiassay_ov . . . . .	7
MultiClass-class . . . . .	8
MultiOmics-class . . . . .	8
plot_chr_distribution . . . . .	9
plot_heatmap . . . . .	10
plot_network . . . . .	11
plot_ridge . . . . .	12
plot_tf_distribution . . . . .	13
plot_venn . . . . .	13
plot_volcano . . . . .	14
run_cnv_integration . . . . .	15
run_genomic_enrich . . . . .	16
run_genomic_integration . . . . .	17
run_met_integration . . . . .	19
run_multiomics . . . . .	20
run_shiny . . . . .	22
run_tf_enrich . . . . .	23
run_tf_integration . . . . .	24

**Index**

**26**

---

gINTomics-package      *gINTomics: Multi-Omics data integration*

---

## Description

gINTomics is an R package for Multi-Omics data integration and visualization. gINTomics is designed to detect the association between the expression of a target and of its regulators, taking into account also their genomics modifications such as Copy Number Variations (CNV) and methylation. What is more, gINTomics allows integration results visualization via a Shiny-based interactive app.

## Author(s)

**Maintainer:** Angelo Velle <angelo.velle@unipd.it> ([ORCID](#))

Authors:

- Francesco Patane' <francesco.patane@unipd.it> ([ORCID](#))
- Chiara Romualdi <chiara.romualdi@unipd.it> ([ORCID](#))

## See Also

Useful links:

- <https://github.com/angelovelle96/gINTomics>
- Report bugs at <https://github.com/angelovelle96/gINTomics/issues>

---

create\_multiassay      *MultiAssayExperiment generation*

---

## Description

This function will generate a proper MultiAssayExperiment suitable for the **run\_multiomics** function.

## Usage

```
create_multiassay(  
  methylation = NULL,  
  cnv_data = NULL,  
  gene_exp = NULL,  
  miRNA_exp = NULL,  
  miRNA_cnv_data = NULL,  
  ...  
)
```

## Arguments

methylation	Matrix or SummarizedExperiment for Methylation data
cnv_data	Matrix or SummarizedExperiment for genes' Copy Number Variation data
gene_exp	Matrix or SummarizedExperiment for Gene expression data
miRNA_exp	Matrix or SummarizedExperiment for miRNA expression data
miRNA_cnv_data	Matrix or SummarizedExperiment for miRNA's Copy Number Variations data
...	Additional arguments to be passed to the function

## Value

A MultiAssayExperiment object containing the provided assays.

## Examples

```
# Example usage:
library(MultiAssayExperiment)
data('mmultiassay_ov')
gene_exp_matrix <- as.matrix(assay(mmultiassay_ov[['gene_exp']]))
miRNA_exp_matrix <- as.matrix(assay(mmultiassay_ov[['miRNA_exp']]))
meth_matrix <- as.matrix(assay(mmultiassay_ov[['methylation']]))
gene_cnv_matrix <- as.matrix(assay(mmultiassay_ov[['cnv_data']]))
miRNA_cnv_matrix <- as.matrix(assay(mmultiassay_ov[['miRNA_cnv_data']]))
create_multiassay(methylation=meth_matrix, cnv_data=gene_cnv_matrix,
  gene_exp=gene_exp_matrix, miRNA_exp=miRNA_exp_matrix,
  miRNA_cnv_data=miRNA_cnv_matrix)
```

---

dot\_plotly

*plotting enrichment*

---

## Description

plotting enrichment

## Usage

```
dot_plotly(
  enrich_result,
  title = NULL,
  showCategory = 10,
  width = 800,
  height = 700
)
```

**Arguments**

enrich\_result    Enrichment analysis results.  
title            Title of the plot.  
showCategory    Number of categories to display.  
width            Width of the plot.  
height           Height of the plot.

**Value**

A plotly object containing the dot plot.

**Examples**

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
#multiomics_integration <- run_multiomics(data = mmultiassay_ov)
#gen_enr <- run_genomic_enrich(multiomics_integration,
#                               qvalueCutoff = 1,
#                               pvalueCutoff = 0.05,
#                               pAdjustMethod = "none")
#dot_plotly(gen_enr, title = "Enrichment Analysis",showCategory = 10)
```

---

extract_model_res	<i>Setting method for extracting results</i>
-------------------	--

---

**Description**

Setting method for extracting results

**Usage**

```
extract_model_res(model_results, ...)

## S4 method for signature 'list'
extract_model_res(
  model_results,
  outliers = TRUE,
  species = "hsa",
  filters = c("hgnc_symbol", "ensembl_gene_id", "entrezgene_id"),
  genes_info = NULL,
  ...
)
```

```

## S4 method for signature 'MultiClass'
extract_model_res(
  model_results,
  outliers = TRUE,
  species = "hsa",
  filters = c("hgnc_symbol", "ensembl_gene_id", "entrezgene_id"),
  genes_info = NULL,
  ...
)

## S4 method for signature 'MultiOmics'
extract_model_res(
  model_results,
  outliers = TRUE,
  species = "hsa",
  filters = c("hgnc_symbol", "ensembl_gene_id", "entrezgene_id"),
  genes_info = NULL,
  ...
)

```

### Arguments

<code>model_results</code>	The model results object from which to extract results.
<code>...</code>	Additional arguments to be passed to specific methods.
<code>outliers</code>	if TRUE (by default), it removes outliers
<code>species</code>	species for the analysis
<code>filters</code>	Specific filters to apply
<code>genes_info</code>	genes info

### Value

A dataframe containing the results of all the integration models provided

### Examples

```

# example code
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
gene_cnv_matrix <- t(as.matrix(assay(mmultiassay_ov[["cnv_data"]])))
gene_exp_matrix <- t(as.matrix(assay(mmultiassay_ov[["gene_exp"]])))
cnv_integration <- run_cnv_integration(
  expression = gene_exp_matrix,
  cnv_data = gene_cnv_matrix
)
data_table <- extract_model_res(cnv_integration)
head(data_table)

```

---

mirna_hsa	<i>miRNA IDs. Dataset containing lastly definition of miRNAs (Names, Accessions, Sequences, Families and others) from different miRBase versions (From miRBase version 6 to version 22).</i>
-----------	--

---

### Description

miRNA IDs. Dataset containing lastly definition of miRNAs (Names, Accessions, Sequences, Families and others) from different miRBase versions (From miRBase version 6 to version 22).

### Usage

```
data(mirna_hsa)
"mirna_hsa"
```

### Value

An object of class `data.frame`.

### Examples

```
# example code
data(mirna_hsa)
head(mirna_hsa)
```

---

mmultiassay_ov	<i>Example data for a standard workflow. This is an example dataset containing a MultiAssayExperiment of 20 ovarian cancer (OVC) patients extracted from the Cancer Genome Atlas (TCGA) database. The object contains all the available input data types: Gene expression data, miRNA expression data, gene methylation data, gene Copy Number Variations and miRNA Copy Number Variations.</i>
----------------	---

---

### Description

Example data for a standard workflow. This is an example dataset containing a MultiAssayExperiment of 20 ovarian cancer (OVC) patients extracted from the Cancer Genome Atlas (TCGA) database. The object contains all the available input data types: Gene expression data, miRNA expression data, gene methylation data, gene Copy Number Variations and miRNA Copy Number Variations.

### Usage

```
data(mmultiassay_ov)
"mmultiomics_ov"
```

**Value**

An object of class [MultiAssayExperiment](#).

**Examples**

```
# example code
data(mmultiassay_ov)
mmultiassay_ov
```

---

MultiClass-class      *MultiClass Class*

---

**Description**

S4 class containing the output of a single integration integration, for which classes has been provided. It's a list in which each element represents the result of the integration for a given class. The length will be equal to the number of classes defined.

**Value**

MultiOmics Class

---

MultiOmics-class      *MultiOmics Class*

---

**Description**

S4 class containing the output of a multiomics integration. It's a list in which each element represents the result of an integration. If all the available omics are provided, it will be a list of integrations: **gene\_genomic\_res**, **mirna\_cnv\_res**, **tf\_res**, **tf\_mirna\_res** and **mirna\_target\_res**

**Value**

MultiOmics Class



---

plot\_chr\_distribution *plotting chr distribution*

---

## Description

plotting chr distribution

## Usage

```
plot_chr_distribution(  
  data_table,  
  class = NULL,  
  omics = NULL,  
  cnv_met = NULL,  
  pval = 0.05  
)
```

## Arguments

data_table	The data table containing information for plotting chromosome distribution.
class	Optional. The class of interactions to include in the plot.
omics	Optional. The type of omics data for the plot.
cnv_met	Optional. The type of copy number variation or methylation data.
pval	Optional. The p-value threshold for significance. Default is 0.05.

## Value

A histogram plot showing chromosome distribution.

## Examples

```
# Example usage:  
library(MultiAssayExperiment)  
data("mmultiassay_ov")  
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])  
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)  
# multiomics_integration <- run_multiomics(data = mmultiassay_ov)  
# data_table <- extract_model_res(multiomics_integration)  
# plot_chr_distribution(data_table, omics = "gene_genomic_res")
```

---

plot_heatmap	<i>plotting heatmap</i>
--------------	-------------------------

---

## Description

plotting heatmap

## Usage

```
plot_heatmap(
  multiomics_integration,
  data_table,
  omics,
  scale = "none",
  genes_number = 50,
  samples_number = 50,
  class = NULL,
  pval = 0.05
)
```

## Arguments

multiomics_integration	The multiomics integration object.
data_table	The data table containing information for the heatmap.
omics	The type of omics data for the heatmap.
scale	Optional. The scale type for the heatmap. Default is "none".
genes_number	Optional. The number of genes to include in the heatmap. Default is 50.
samples_number	Number of samples to include in the heatmap. If this number is inferior to the total number of samples, the n most variable samples will be selected
class	Optional. The class of interactions to include in the heatmap.
pval	Optional. The p-value threshold for significance in the heatmap. Default is 0.05.

## Value

A heatmap plot.

## Examples

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
# multiomics_integration <- run_multiomics(data = mmultiassay_ov)
# data_table <- extract_model_res(multiomics_integration)
```

```
# data_table <- data_table[!is.na(data_table$cnv_met),]  
# plot_heatmap(multiomics_integration, data_table, omics = "gene_genomic_res")
```

---

plot_network	<i>Plotting network</i>
--------------	-------------------------

---

## Description

Plotting network

## Usage

```
plot_network(data_table, num_interactions = 300, class = NULL, pval = 0.05)
```

## Arguments

data_table	The data table containing network information.
num_interactions	The number of interactions to display in the network (default: 300).
class	Optional. The class of interactions to include in the plot.
pval	The p-value threshold for selecting interactions (default: 0.05).

## Value

A network plot.

## Examples

```
# Example usage:  
library(MultiAssayExperiment)  
data("mmultiassay_ov")  
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])  
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)  
# multiomics_integration <- run_multiomics(data = mmultiassay_ov)  
# data_table <- extract_model_res(multiomics_integration)  
# plot_network(data_table)
```

---

plot_ridge	<i>plotting ridge</i>
------------	-----------------------

---

### Description

plotting ridge

### Usage

```
plot_ridge(data_table, class = NULL, omics = NULL, cnv_met = NULL)
```

### Arguments

data_table	The data table containing information for the ridge plot.
class	Optional. The class of interactions to include in the ridge plot.
omics	Optional. The omics type for the ridge plot.
cnv_met	Optional. Indicates whether the ridge plot is for CNV or MET omics (only applicable if omics is specified).

### Value

A ridge plot.

### Examples

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
gene_cnv_matrix <- t(as.matrix(assay(mmultiassay_ov[["cnv_data"]]))))
gene_exp_matrix <- t(as.matrix(assay(mmultiassay_ov[["gene_exp"]]))))
cnv_integration <- run_cnv_integration(
  expression = gene_exp_matrix,
  cnv_data = gene_cnv_matrix
)
data_table <- extract_model_res(cnv_integration)
data_table <- data_table[data_table$cov!="(Intercept)",]
plot_ridge(data_table)
```

---

plot\_tf\_distribution    *plotting TF distribution*

---

**Description**

plotting TF distribution

**Usage**

```
plot_tf_distribution(data_table, class = NULL, pval = 0.05)
```

**Arguments**

data_table	The data table containing TF information.
class	Optional. The class of interactions to include in the distribution plot.
pval	Optional. The p-value threshold for significance in the distribution plot. Default is 0.05.

**Value**

A TF distribution plot.

**Examples**

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
# multiomics_integration <- run_multiomics(data = mmultiassay_ov)
# data_table <- extract_model_res(multiomics_integration)
# plot_tf_distribution(data_table, pval=0.5)
```

---

plot\_venn                    *plotting venn*

---

**Description**

plotting venn

**Usage**

```
plot_venn(data_table, class = NULL)
```

**Arguments**

`data_table` The data table containing information for the Venn diagram.  
`class` Optional. The class of interactions to include in the Venn diagram.

**Value**

A Venn diagram plot.

**Examples**

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
# multiomics_integration <- run_multiomics(data = mmultiassay_ov)
# data_table <- extract_model_res(multiomics_integration)
# plot_venn(data_table)
```

---

plot\_volcano

*plotting volcano*

---

**Description**

plotting volcano

**Usage**

```
plot_volcano(data_table, class = NULL, omics = NULL, cnv_met = NULL)
```

**Arguments**

`data_table` The data table containing information for the volcano plot.  
`class` Optional. The class of interactions to include in the volcano plot.  
`omics` Optional. The omics type for the volcano plot.  
`cnv_met` Optional. Indicates whether the volcano plot is for CNV or MET omics (only applicable if omics is specified).

**Value**

A volcano plot.

## Examples

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
multiomics_integration <- run_multiomics(data = mmultiassay_ov)
data_table <- extract_model_res(multiomics_integration)
plot_volcano(data_table, omics = "gene_genomic_res", cnv_met = "cnv")
```

---

run\_cnv\_integration     *Integration of expression and Copy Number Variations*

---

## Description

This function will perform an integration of expression data and Copy Number Variations data

## Usage

```
run_cnv_integration(
  expression,
  cnv_data,
  sequencing_data = TRUE,
  normalize = TRUE,
  norm_method = "TMM",
  class = NULL,
  run_deg = TRUE,
  BPPARAM = SerialParam(),
  ...
)
```

## Arguments

expression	Matrix or data.frame containing the expression values for each model. Rows represent samples, while each column represents the different response variables of the models.
cnv_data	Matrix or data.frame containing the Copy Number variation status for the models. Rows represent samples, while columns represent the different covariates. If <b>interactions</b> are not provided, they will be automatically generated and for each gene contained in <b>expression</b> the model will look for the same gene in <b>cnv_data</b>
sequencing_data	logical. Are expression data obtained from RNA sequencing ? Default is set to TRUE
normalize	logical. Should expression data be normalized ? Default is set to TRUE
norm_method	Normalization method to be used for expression data. One of "TMM" (default), "TMMwsp", "RLE", "upperquartile", "none".

class	Character vector specifying the classes for differential expression analysis.
run_deg	Logical. Should differential expression analysis be performed? Default is set to TRUE.
BPPARAM	A BiocParallelParam object specifying the parallel backend to be used.
...	Additional arguments to be passed to internal functions.

**Value**

A list or a [MultiClass](#) object if **class** is provided containing the results of the CNV integration

**Examples**

```
# Example usage_multi:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
gene_cnv_matrix <- t(as.matrix(assay(mmultiassay_ov[["cnv_data"]])))
gene_exp_matrix <- t(as.matrix(assay(mmultiassay_ov[["gene_exp"]])))
cnv_integration <- run_cnv_integration(
  expression = gene_exp_matrix,
  cnv_data = gene_cnv_matrix
)
```

---

run\_genomic\_enrich      *Running genomic enrichment analysis*

---

**Description**

Running genomic enrichment analysis

**Usage**

```
run_genomic_enrich(
  model_results,
  species = "hsa",
  pvalueCutoff = 0.1,
  pAdjustMethod = "BH",
  qvalueCutoff = 0.1,
  ont = "all",
  BPPARAM = BiocParallel::SerialParam(),
  extracted_data = NULL,
  ...
)
```



**Arguments**

model_results	Model integration results, typically a list containing different types of genomic results
species	Species to select for the enrichment analysis. Default is 'hsa' (Homo sapiens).
pvalueCutoff	P-value cutoff for significant enrichment. Default is 0.1.
pAdjustMethod	Method for adjusting p-values. Default is 'BH' (Benjamini & Hochberg).
qvalueCutoff	Q-value cutoff for significant enrichment. Default is 0.1.
ont	Ontology to use for the enrichment analysis. Default is 'all'.
BPPARAM	A BiocParallelParam object specifying parallelization options. Default is BiocParallel::SerialParam().
extracted_data	Pre-extracted data for enrichment analysis. If NULL, function will extract relevant data from model_results.
...	Additional arguments to be passed to the internal enrichment function.

**Value**

A list containing enrichment results. If CNV and methylation data are available, it returns a nested list with results for each data type.

**Examples**

```
# Example usage:
library(MultiAssayExperiment)
data(mmultiassay_ov)
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:200,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
#multiomics_integration <- run_multiomics(mmultiassay_ov)
#gen_enr <- run_genomic_enrich(multiomics_integration, qvalueCutoff = 1,
#pvalueCutoff = 0.05, pAdjustMethod = 'none')
```

---

run\_genomic\_integration

*Integration of expression, Copy Number Variations and methylation data*

---

**Description**

This function will perform an integration of expression data and Copy Number Variations data

**Usage**

```
run_genomic_integration(
  expression,
  cnv_data,
  methylation,
  sequencing_data = TRUE,
  normalize = TRUE,
  norm_method = "TMM",
  interactions = NULL,
  class = NULL,
  scale = TRUE,
  run_deg = TRUE,
  BPPARAM = SerialParam(),
  ...
)
```

**Arguments**

expression	Matrix or data.frame containing the expression values for each model. Rows represent samples, while each column represents the different response variables of the models.
cnv_data	Matrix or data.frame containing the Copy Number variation status for the models. Rows represent samples, while columns represent the different covariates. If <b>interactions</b> are not provided, they will be automatically generated and for each gene contained in <b>expression</b> the model will look for the same gene in <b>cnv_data</b>
methylation	Matrix or data.frame containing the methylation values for the models. Rows represent samples, while columns represent the different covariates. If <b>interactions</b> are not provided, they will be automatically generated and for each gene contained in <b>expression</b> the model will look for the same gene in <b>methylation</b>
sequencing_data	logical. Are expression data obtained from RNA sequencing ? Default is set to TRUE
normalize	logical. Should expression data be normalized ? Default is set to TRUE
norm_method	Normalization method to be used for expression data. One of "TMM" (default), "TMMwsp", "RLE", "upperquartile", "none".
interactions	A list of character vectors containing the interactions between response variable and covariates. The names of the list should match the response variables while the character contained in each element of the list should match the covariates. If NULL (default), the interactions will be automatically defined according to response variable's colnames.
class	Character vector specifying the classes for differential expression analysis.
scale	Logical. Should the data be scaled? Default is set to TRUE.
run_deg	Logical. Should differential expression analysis be performed? Default is set to TRUE.

BPPARAM        A BiocParallelParam object specifying the parallel backend to be used.  
 ...            Additional arguments to be passed to internal functions.

### Value

A list or a [MultiClass](#) object if **class** is provided containing the results of the Genomic integration

### Examples

```
# Example usage_multi:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
meth_matrix <- t(as.matrix(assay(mmultiassay_ov[["methylation"]])))
gene_exp_matrix <- t(as.matrix(assay(mmultiassay_ov[["gene_exp"]])))
gene_cnv_matrix <- t(as.matrix(assay(mmultiassay_ov[["cnv_data"]])))
genomic_integration <- run_genomic_integration(
  expression = gene_exp_matrix,
  cnv_data = gene_cnv_matrix, methylation = meth_matrix
)
```

---

run\_met\_integration    *Integration of expression and methylation*

---

### Description

This function will perform an integration of expression data and methylation data

### Usage

```
run_met_integration(
  expression,
  methylation,
  sequencing_data = TRUE,
  normalize = TRUE,
  norm_method = "TMM",
  class = NULL,
  run_deg = TRUE,
  BPPARAM = SerialParam(),
  ...
)
```

### Arguments

expression        Matrix or data.frame containing the expression values for each model. Rows represent samples, while each column represents the different response variables of the models.

methylation	Matrix or data.frame containing the methylation values for the models. Rows represent samples, while columns represent the different covariates. If <b>interactions</b> are not provided, they will be automatically generated and for each gene contained in <b>expression</b> the model will look for the same gene in <b>methylation</b>
sequencing_data	logical. Are expression data obtained from RNA sequencing ? Default is set to TRUE
normalize	logical. Should expression data be normalized ? Default is set to TRUE
norm_method	Normalization method to be used for expression data. One of "TMM" (default), "TMMwsp", "RLE", "upperquartile", "none".
class	Character vector specifying the classes for differential expression analysis.
run_deg	Logical. Should differential expression analysis be performed? Default is set to TRUE.
BPPARAM	A BiocParallelParam object specifying the parallel backend to be used.
...	Additional arguments to be passed to internal functions.

### Value

A list or a [MultiClass](#) object if **class** is provided containing the results of the Methylation integration

### Examples

```
# Example usage_multi:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
meth_matrix <- t(as.matrix(assay(mmultiassay_ov[["methylation"]])))
gene_exp_matrix <- t(as.matrix(assay(mmultiassay_ov[["gene_exp"]])))
met_integration <- run_met_integration(
  expression = gene_exp_matrix,
  methylation = meth_matrix
)
```

---

run\_multiomics

*Complete Multi-Omics integration*

---

### Description

This function will perform a complete Multi-Omics integration on a MultiAssayExperiment

**Usage**

```
run_multiomics(
  data,
  interactions_met = NULL,
  interactions_miRNA_target = NULL,
  interactions_tf = NULL,
  interactions_tf_miRNA = NULL,
  RNAseq = TRUE,
  miRNAseq = TRUE,
  normalize_miRNA_expr = TRUE,
  normalize_gene_expr = TRUE,
  norm_method_gene_expr = "TMM",
  norm_method_miRNA_expr = "TMM",
  class = NULL,
  BPPARAM = SerialParam()
)
```

**Arguments**

data	A MultiAssayExperiment. It can be generated exploiting the <b>generate_multiassay</b> function.
interactions_met	<b>interactions</b> as for <b>run_met_integration</b>
interactions_miRNA_target	miRNA-target interactions as requested by <b>run_tf_integration</b>
interactions_tf	TF-target interactions as requested by <b>run_tf_integration</b>
interactions_tf_miRNA	TF-target interactions as requested by <b>run_tf_integration</b>
RNAseq	logical. Are gene expression data obtained from RNA sequencing ? Default is set to TRUE
miRNAseq	logical. Are miRNA expression data obtained from miRNA sequencing ? Default is set to TRUE
normalize_miRNA_expr	logical. Should miRNA expression data be normalized ? Default is set to TRUE
normalize_gene_expr	logical. Should gene expression data be normalized ? Default is set to TRUE
norm_method_gene_expr	Normalization method to be used for gene expression data. One of "TMM" (default), "TMMwsp", "RLE", "upperquartile", "none".
norm_method_miRNA_expr	Normalization method to be used for miRNA expression data. One of "TMM" (default), "TMMwsp", "RLE", "upperquartile", "none".
class	Character vector specifying the classes for differential expression analysis.
BPPARAM	A BiocParallelParam object specifying the parallel backend to be used.

**Value**

A [MultiOmics](#) object containing the results of all the possible integration models

**Examples**

```
# Example usage_multiomics:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
multiomics_integration <- run_multiomics(data = mmultiassay_ov)
```

---

run\_shiny

*Start a Shiny application for integrated multi-omics data analysis.*

---

**Description**

The `run_shiny` function launches an interactive Shiny application that allows users to explore and analyze integrated multi-omics data through various visualizations and analyses.

**Usage**

```
run_shiny(multiomics_integration)
```

**Arguments**

`multiomics_integration`

An object representing the integration of multi-omics data, compatible with the [extract\\_model\\_res](#) function.

**Details**

The `run_shiny` function extracts model results from `multiomics_integration`, performs preprocessing operations to prepare the data for the Shiny user interface, creates the user interface and server for the Shiny application.

**Value**

No return value. The function starts an interactive Shiny application.

**References**

Description of the multi-omics data model and integrated analysis techniques used.

**See Also**

[extract\\_model\\_res](#)

## Examples

```
# Example usage:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
# multiomics_integration <- run_multiomics(data = mmultiassay_ov)
# app <- run_shiny(multiomics_integration)
```

---

run_tf_enrich	<i>Running TF enrichment analysis</i>
---------------	---------------------------------------

---

## Description

Running TF enrichment analysis

## Usage

```
run_tf_enrich(
  model_results,
  species = "hsa",
  pvalueCutoff = 0.1,
  qvalueCutoff = 0.1,
  pAdjustMethod = "BH",
  ont = "all",
  BPPARAM = BiocParallel::SerialParam(),
  extracted_data = NULL,
  ...
)
```

## Arguments

model_results	Model integration results, typically a list containing TF data.
species	Species to select for the enrichment analysis. Default is 'hsa' (Homo sapiens).
pvalueCutoff	P-value cutoff for significant enrichment. Default is 0.1.
qvalueCutoff	Q-value cutoff for significant enrichment. Default is 0.1.
pAdjustMethod	Method for adjusting p-values. Default is 'BH' (Benjamini & Hochberg).
ont	Ontology to use for the enrichment analysis. Default is 'all'.
BPPARAM	A BiocParallelParam object specifying parallelization options. Default is BiocParallel::SerialParam().
extracted_data	Pre-extracted data for enrichment analysis. If NULL, function will extract relevant data from model_results.
...	Additional arguments to be passed to the internal enrichment function.

**Value**

A list containing TF enrichment results.

**Examples**

```
# Example usage:
library(MultiAssayExperiment)
data(mmultiassay_ov)
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:200,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
#multiomics_integration <- run_multiomics(mmultiassay_ov)
#run_tf_enrich(multiomics_integration, qvalueCutoff = 1, pvalueCutoff = 0.05,
#pAdjustMethod = 'none')
```

---

run_tf_integration	<i>Integration of expression and Transcription Factors / Generic Regulators</i>
--------------------	---

---

**Description**

This function will perform an integration of gene/miRNA expression data and Transcription Factors expression. Moreover, every type of regulator can be provided to the function as covariate through the **tf\_expression** argument. Interactions for TF-target, miRNA-target and TF-miRNA integration will be automatically downloaded by the function as defined by the **type** argument. Other types of interactions should be provided through the **interactions** argument.

**Usage**

```
run_tf_integration(  
  expression,  
  tf_expression = expression,  
  interactions = NULL,  
  type = "none",  
  sequencing_data = TRUE,  
  species = "hsa",  
  normalize = TRUE,  
  norm_method = "TMM",  
  normalize_cov = TRUE,  
  norm_method_cov = "TMM",  
  class = NULL,  
  run_deg = TRUE,  
  BPPARAM = SerialParam(),  
  ...  
)
```



**Arguments**

expression	Matrix or data.frame containing the expression values for each model. Rows represent samples, while each column represents the different response variables of the models.
tf_expression	Matrix or data.frame containing the expression values for the models. Rows represent samples, while columns represent the different covariates. If not provided, it will be set equal to <b>expression</b> .
interactions	A list of character vectors containing the interactions between response variable and covariates. The names of the list should match the response variables while the character contained in each element of the list should match the covariates. If NULL (default), the interactions will be automatically downloaded according to the <b>type</b> argument.
type	A character defining the type of regulation under analysis. Should be one of "tf_miRNA", "tf", "miRNA_target".
sequencing_data	logical. Are expression data obtained from RNA sequencing ? Default is set to TRUE
species	species information for interactions download. Fully supported species are "hsa" (default) and "mmu".
normalize	logical. Should expression data be normalized ? Default is set to TRUE
norm_method	Normalization method to be used for expression data. One of "TMM" (default), "TMMwsp", "RLE", "upperquartile", "none".
normalize_cov	Same as <b>normalize</b> but for covariates.
norm_method_cov	Same as <b>norm_method</b> but for covariates.
class	Character vector specifying the classes for differential expression analysis.
run_deg	Logical. Should differential expression analysis be performed? Default is set to TRUE.
BPPARAM	A BiocParallelParam object specifying the parallel backend to be used.
...	Additional arguments to be passed to internal functions.

**Value**

A list or a [MultiClass](#) object if **class** is provided containing the results of the transcriptional integration

**Examples**

```
# Example usage_multi:
library(MultiAssayExperiment)
data("mmultiassay_ov")
tmp <- lapply(experiments(mmultiassay_ov), function(x) x[1:20,])
mmultiassay_ov <- MultiAssayExperiment(experiments = tmp)
gene_exp_matrix <- t(as.matrix(assay(mmultiassay_ov[["gene_exp"]]))))
tf_integration <- run_tf_integration(expression = gene_exp_matrix, type="tf")
```

# Index

- \* **Data analysis**
    - run\_shiny, [22](#)
  - \* **Function**
    - run\_shiny, [22](#)
  - \* **Integration**
    - run\_shiny, [22](#)
  - \* **Interactive**
    - run\_shiny, [22](#)
  - \* **Multi-omics**
    - run\_shiny, [22](#)
  - \* **Shiny**
    - run\_shiny, [22](#)
  - \* **Visualization**
    - run\_shiny, [22](#)
  - \* **analysis**
    - run\_shiny, [22](#)
  - \* **integration**
    - run\_shiny, [22](#)
  - \* **internal**
    - gINTomics-package, [3](#)
  - \* **multiomics**
    - run\_shiny, [22](#)
  - \* **shiny**
    - run\_shiny, [22](#)
  - \* **visualization**
    - run\_shiny, [22](#)
- create\_multiassay, [3](#)
- data.frame, [7](#)
- dot\_plotly, [4](#)
- extract\_model\_res, [5](#), [22](#)
- extract\_model\_res, list-method  
(extract\_model\_res), [5](#)
- extract\_model\_res, MultiClass-method  
(extract\_model\_res), [5](#)
- extract\_model\_res, MultiOmics-method  
(extract\_model\_res), [5](#)
- gINTomics (gINTomics-package), [3](#)
- gINTomics-package, [3](#)
- mirna\_hsa, [7](#)
- mmultiassay\_ov, [7](#)
- MultiAssayExperiment, [8](#)
- MultiClass, [16](#), [19](#), [20](#), [25](#)
- MultiClass-class, [8](#)
- MultiOmics, [22](#)
- MultiOmics-class, [8](#)
- plot\_chr\_distribution, [9](#)
- plot\_heatmap, [10](#)
- plot\_network, [11](#)
- plot\_ridge, [12](#)
- plot\_tf\_distribution, [13](#)
- plot\_venn, [13](#)
- plot\_volcano, [14](#)
- run\_cnv\_integration, [15](#)
- run\_genomic\_enrich, [16](#)
- run\_genomic\_integration, [17](#)
- run\_met\_integration, [19](#)
- run\_multiomics, [20](#)
- run\_shiny, [22](#)
- run\_tf\_enrich, [23](#)
- run\_tf\_integration, [24](#)