

# Package ‘biobroom’

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**Title** Turn Bioconductor objects into tidy data frames

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**Author** Andrew J. Bass, David G. Robinson, Steve Lianoglou, Emily Nelson, John D. Storey, with contributions from Laurent Gatto

**Maintainer** John D. Storey <jstorey@princeton.edu> and Andrew J. Bass <ajbass@emory.edu>

**Description** This package contains methods for converting standard objects constructed by bioinformatics packages, especially those in Bioconductor, and converting them to tidy data. It thus serves as a complement to the broom package, and follows the same the tidy, augment, glance division of tidying methods. Tidying data makes it easy to recombine, reshape and visualize bioinformatics analyses.

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**Depends** R (>= 3.0.0), broom

**License** LGPL

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## Contents

augment_sva . . . . .	2
biobroom . . . . .	3
DESeq2_tidiers . . . . .	3
edgeR_tidiers . . . . .	5
ExpressionSet_tidiers . . . . .	6
hammer . . . . .	7
limma_tidiers . . . . .	8
list_tidiers . . . . .	10
MSnSet_tidiers . . . . .	11
qvalue_tidiers . . . . .	12
SummarizedExperiment_tidiers . . . . .	14
tidy.deSet . . . . .	15
tidy.GRanges . . . . .	16
<b>Index</b>	<b>18</b>

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augment_sva	<i>Tidying methods for a sva list</i>
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### Description

These are methods for turning a sva list, from the sva package, into a tidy data frame. `tidy` returns a data.frame of the estimated surrogate variables, `glance` returns a data.frame of the posterior probabilities, and `glance` returns a data.frame with only the number of surrogate variables.

### Usage

```
augment_sva(x, data, ...)

tidy_sva(x, addVar = NULL, ...)

glance_sva(x, ...)
```

### Arguments

x	sva list
data	Original data
...	extra arguments (not used)
addVar	add additional coefficients to the estimated surrogate variables

**Value**

All tidying methods return a `data.frame` without rownames. The structure depends on the method chosen.

`augment` returns one row per gene. It always contains the columns

`pprob.gam`      Posterior probability each gene is affected by heterogeneity

`pprob.b`        Posterior probability each gene is affected by model

`tidy` returns the estimate surrogate variables.

`glance` returns the estimate surrogate variables.

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biobroom	<i>Convert Bioconductor Object into Tidy Data Frames</i>
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**Description**

This package contains methods for converting standard objects constructed by bioinformatics packages, especially those in Bioconductor, and converting them to tidy data. It thus serves as a complement to the broom package, and follows the same the tidy, augment, glance division of tidying methods. Tidying data makes it easy to recombine, reshape and visualize bioinformatics analyses.

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DESeq2_tidiers	<i>Tidying methods for DESeq2 DESeqDataSet objects</i>
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---

**Description**

This reshapes a DESeq2 expressionset object into a tidy format. If the dataset contains hypothesis test results (p-values and estimates), this summarizes one row per gene per possible contrast.

**Usage**

```
## S3 method for class 'DESeqDataSet'
tidy(x, colData = FALSE, intercept = FALSE, ...)
```

```
## S3 method for class 'DESeqResults'
tidy(x, ...)
```

**Arguments**

<code>x</code>	DESeqDataSet object
<code>colData</code>	whether <code>colData</code> should be included in the tidied output for those in the DESeqDataSet object. If dataset includes hypothesis test results, this is ignored
<code>intercept</code>	whether to include hypothesis test results from the (Intercept) term. If dataset does not include hypothesis testing, this is ignored
<code>...</code>	extra arguments (not used)

**Details**

colDat=TRUE adds covariates from colData to the data frame.

**Value**

If the dataset contains results (p-values and log2 fold changes), the result is a data frame with the columns

term	The contrast being tested, as given to results
gene	gene ID
baseMean	mean abundance level
estimate	estimated log2 fold change
stderror	standard error in log2 fold change estimate
statistic	test statistic
p.value	p-value
p.adjusted	adjusted p-value

If the dataset does not contain results (DESeq has not been run on it), tidy defaults to tidying the counts in the dataset:

gene	gene ID
sample	sample ID
count	number of reads in this gene in this sample

If colData = TRUE, it also merges this with the columns present in colData(x).

**Examples**

```
# From DESeq2 documentation

if (require("DESeq2")) {
  dds <- makeExampleDESeqDataSet(betaSD = 1)

  tidy(dds)
  # With design included
  tidy(dds, colData=TRUE)

  # add a noise confounding effect
  colData(dds)$noise <- rnorm(nrow(colData(dds)))
  design(dds) <- (~ condition + noise)

  # perform differential expression tests
  ddsres <- DESeq(dds, test = "Wald")
  # now results are per-gene, per-term
  tidied <- tidy(ddsres)
  tidied

  if (require("ggplot2")) {
    ggplot(tidied, aes(p.value)) + geom_histogram(binwidth = .05) +
```

```

    facet_wrap(~ term, scale = "free_y")
  }
}

```

---

edgeR\_tidiers

*Tidiers for edgeR's differential expression objects*


---

## Description

Tidy, augment and glance methods for turning edgeR objects into tidy data frames, where each row represents one observation and each column represents one column.

## Usage

```

## S3 method for class 'DGEEExact'
tidy(x, ...)

## S3 method for class 'DGEList'
tidy(x, addSamples = FALSE, ...)

## S3 method for class 'DGEList'
augment(x, data = NULL, ...)

## S3 method for class 'DGEEExact'
glance(x, alpha = 0.05, p.adjust.method = "fdr",
       ...)

```

## Arguments

x	DGEEExact, DGEList object
...	extra arguments (not used)
addSamples	Merge information from samples. Default is FALSE.
data	merge data to augment. This is particularly useful when merging gene names or other per-gene information. Default is NULL.
alpha	Confidence level to test for significance
p.adjust.method	Method for adjusting p-values to determine significance; can be any in p.adjust.methods

## Value

tidy defaults to tidying the counts in the dataset:

gene	gene ID
sample	sample ID
count	number of reads in this gene in this sample

If `addSamples = TRUE`, it also merges this with the sample information present in `x$samples`.

`augment` returns per-gene information (DGEList only)

`glance` returns one row with the columns (DGEEExact only)

`significant`      number of significant genes using desired adjustment method and confidence level

`comparison`      The pair of groups compared by edgeR, delimited by /

### Examples

```
if (require("edgeR")) {
  library(Biobase)
  data(hammer)
  hammer.counts <- exprs(hammer)[, 1:4]
  hammer.treatment <- phenoData(hammer)$protocol[1:4]

  y <- DGEList(counts=hammer.counts,group=hammer.treatment)
  y <- calcNormFactors(y)
  y <- estimateCommonDisp(y)
  y <- estimateTagwiseDisp(y)
  et <- exactTest(y)

  head(tidy(et))
  head(glance(et))
}
```

---

ExpressionSet\_tidiers *Tidying methods for Biobase's ExpressionSet objects*

---

### Description

Tidying methods for Biobase's ExpressionSet objects

### Usage

```
## S3 method for class 'ExpressionSet'
tidy(x, addPheno = FALSE,
     assay = Biobase::assayDataElementNames(x)[1L], ...)
```

### Arguments

<code>x</code>	ExpressionSet object
<code>addPheno</code>	whether columns should be included in the tidied output for those in the ExpressionSet's <code>phenoData</code>
<code>assay</code>	The name of the <a href="#">assayDataElement</a> to use as the values to tidy. Defaults to <code>assayDataElementNames(x)[1L]</code> , which is usually equivalent to <code>exprs(x)</code> .
<code>...</code>	extra arguments (not used)

## Details

addPheno=TRUE adds columns that are redundant (since they add per-sample information to a per-sample-per-gene data frame), but that are useful for some kinds of graphs and analyses.

## Value

tidy returns a data frame with one row per gene-sample combination, with columns

gene	gene name
sample	sample name (from column names)
value	expressions on log2 scale

## Examples

```
library(Biobase)
# import ExpressionSet object
data(hammer)

# Use tidy to extract genes, sample ids and measured value
tidy(hammer)
# add phenoType data
tidy(hammer, addPheno=TRUE)
```

---

hammer

*ExpressionSet results from Hammer et al 2010*

---

## Description

An ExpressionSet containing the results of the Hammer et al 2010 RNA-Seq study on the nervous system of rats (Hammer et al 2010).

This was downloaded from the ReCount database of analysis-ready RNA-Seq datasets (Frazee et al 2011).

Hammer, P., Banck, M. S., Amberg, R., Wang, C., Petznick, G., Luo, S., Khrebtukova, I., Schroth, G. P., Beyerlein, P., and Beutler, A. S. (2010). mRNA-seq with agnostic splice site discovery for nervous system transcriptomics tested in chronic pain. *Genome research*, 20(6), 847-860. <http://www.ncbi.nlm.nih.gov/pmc/articles/PMC2877581/>

Frazee, A. C., Langmead, B., and Leek, J. T. (2011). ReCount: a multi-experiment resource of analysis-ready RNA-seq gene count datasets. *BMC Bioinformatics*, 12, 449. <http://bowtie-bio.sourceforge.net/recount/>

## Usage

```
hammer
```

**Format**

An object of class ExpressionSet with 29516 rows and 8 columns.

**Value**

ExpressionSet

---

limma_tidiers	<i>Tidiers for the output of limma (linear models for microarray analysis)</i>
---------------	--------------------------------------------------------------------------------

---

**Description**

Tidy, augment, and glance methods for MArrayLM objects, which contain the results of gene-wise linear models to microarray datasets. This class is the output of the lmFit and eBayes functions.

**Usage**

```
## S3 method for class 'MArrayLM'
tidy(x, intercept = FALSE, ...)
```

```
## S3 method for class 'MArrayLM'
augment(x, data, ...)
```

```
## S3 method for class 'MArrayLM'
glance(x, ...)
```

```
## S3 method for class 'MAlist'
tidy(x, ...)
```

```
## S3 method for class 'EList'
tidy(x, addTargets = FALSE, ...)
```

**Arguments**

x	MArrayLM, MAlist, Elist object
intercept	whether the (Intercept) term should be included (default FALSE)
...	extra arguments, not used
data	original expression matrix; if missing, augment returns only the computed per-gene statistics
addTargets	Add sample level information. Default is FALSE.

**Details**

Tidying this fit computes one row per coefficient per gene, while augmenting returns one row per gene, with per-gene statistics included. (This is thus a rare case where the augment output has more rows than the tidy output. This is a side effect of the fact that the input to limma is not tidy but rather a one-row-per-gene matrix).



**Value**

The output of tidying functions is always a data frame without rownames.

`tidy` returns one row per gene per coefficient. It always contains the columns

<code>gene</code>	The name of the gene (extracted from the rownames of the input matrix)
<code>term</code>	The coefficient being estimated
<code>estimate</code>	The estimate of each per-gene coefficient

Depending on whether the object comes from eBayes, it may also contain

<code>statistic</code>	Empirical Bayes t-statistic
<code>p.value</code>	p-value computed from t-statistic
<code>lod</code>	log-of-odds score

`augment` returns one row per gene, containing the original gene expression matrix if provided. It then adds columns containing the per-gene statistics included in the MArrayLM object, each prepended with a `.`:

<code>.gene</code>	gene ID, obtained from the rownames of the input
<code>.sigma</code>	per-gene residual standard deviation
<code>.df.residual</code>	per-gene residual degrees of freedom

The following columns may also be included, depending on which have been added by `lmFit` and `eBayes`:

<code>.AMean</code>	average intensity across probes
<code>.statistic</code>	moderated F-statistic
<code>.p.value</code>	p-value generated from moderated F-statistic
<code>.df.total</code>	total degrees of freedom per gene
<code>.df.residual</code>	residual degrees of freedom per gene
<code>.s2.prior</code>	prior estimate of residual variance
<code>.s2.post</code>	posterior estimate of residual variance

`glance` returns one row, containing

<code>rank</code>	rank of design matrix
<code>df.prior</code>	empirical Bayesian prior degrees of freedom
<code>s2.prior</code>	empirical Bayesian prior residual standard deviation

`tidy` returns a data frame with one row per gene-sample combination, with columns

<code>gene</code>	gene name
<code>sample</code>	sample name (from column names)
<code>value</code>	expressions on log <sub>2</sub> scale

`tidy` returns a data frame with one row per gene-sample combination, with columns

gene	gene name
sample	sample name (from column names)
value	expressions on log2 scale
weight	present if weights is set
other columns	if present and if addTargets is set

### Examples

```

if (require("limma")) {
  # create random data and design
  set.seed(2014)
  dat <- matrix(rnorm(1000), ncol=4)
  dat[, 1:2] <- dat[, 1:2] + .5 # add an effect
  rownames(dat) <- paste0("g", 1:nrow(dat))
  des <- data.frame(treatment = c("a", "a", "b", "b"),
                    confounding = rnorm(4))

  lfit <- lmFit(dat, model.matrix(~ treatment + confounding, des))
  eb <- eBayes(lfit)
  head(tidy(lfit))
  head(tidy(eb))

  if (require("ggplot2")) {
    # the tidied form puts it in an ideal form for plotting
    ggplot(tidy(lfit), aes(estimate)) + geom_histogram(binwidth=1) +
      facet_wrap(~ term)
    ggplot(tidy(eb), aes(p.value)) + geom_histogram(binwidth=.2) +
      facet_wrap(~ term)
  }
}

```

---

list\_tidiers

*Tidiers for return values from functions that aren't S3 objects*

---

### Description

This method handles the return values of functions that return lists rather than S3 objects, such as `sva`, and therefore cannot be handled by S3 dispatch.

### Usage

```

## S3 method for class 'list'
tidy(x, ...)

## S3 method for class 'list'
glance(x, ...)

```

**Arguments**

x	list object
...	extra arguments, passed to the tidying function

**Details**

Those tidiers themselves are implemented as functions of the form tidy\_<function> that are not exported.

---

MSnSet_tidiers	<i>Tidying methods for Biobase's ExpressionSet objects</i>
----------------	------------------------------------------------------------

---

**Description**

Tidying methods for Biobase's ExpressionSet objects

**Usage**

```
## S3 method for class 'MSnSet'
tidy(x, addPheno = FALSE, ...)
```

**Arguments**

x	MSnSet object
addPheno	whether columns should be included in the tidied output for those in the MSnSet's phenoData
...	extra arguments (not used)

**Details**

addPheno=TRUE adds columns that are redundant (since they add per-sample information to a per-sample-per-gene data frame), but that are useful for some kinds of graphs and analyses.

**Value**

tidy returns a data frame with one row per gene-sample combination, with columns

protein	protein name
sample	sample name (from column names)
value	protein quantitation data

**Examples**

```

if (require("MSnbase")) {
  library(MSnbase)
  # import MSnSet object
  data(msnset)

  # Use tidy to extract genes, sample ids and measured value
  tidy(msnset)
  # add phenoType data
  tidy(msnset, addPheno=TRUE)
}

```

---

qvalue\_tidiers

*Tidying methods for a qvalue object*


---

**Description**

These are methods for turning a qvalue object, from the qvalue package for false discovery rate control, into a tidy data frame. `augment` returns a data.frame of the original p-values combined with the computed q-values and local false discovery rates, `tidy` constructs a table showing how the estimate of  $\pi_0$  (the proportion of true nulls) depends on the choice of the tuning parameter  $\lambda$ , and `glance` returns a data.frame with only the chosen  $\pi_0$  value.

**Usage**

```

## S3 method for class 'qvalue'
tidy(x, ...)

## S3 method for class 'qvalue'
augment(x, data, ...)

## S3 method for class 'qvalue'
glance(x, ...)

```

**Arguments**

x	qvalue object
...	extra arguments (not used)
data	Original data

**Value**

All tidying methods return a data.frame without rownames. The structure depends on the method chosen.

`tidy` returns one row for each choice of the tuning parameter  $\lambda$  that was considered (argument `lambda` to `qvalue`), containing

lambda	the tuning parameter
pi0	corresponding estimate of pi0
smoothed	whether the estimate has been spline-smoothed)

If `pi0.method="smooth"`, the `pi0` estimates and smoothed values both appear in the table. If `pi0.method="bootstrap"`, `smoothed` is `FALSE` for all entries.

`augment` returns a `data.frame` with

<code>p.value</code>	the original p-values given to <code>qvalue</code>
<code>q.value</code>	the computed q-values
<code>lfdr</code>	the local false discovery rate

`glance` returns a one-row `data.frame` containing

<code>pi0</code>	the estimated pi0 (proportion of nulls)
<code>lambda</code>	lambda used to compute pi0. Note that if pi0 is 1, this may be NA since it can be ambiguous which lambda was used

## Examples

```
library(ggplot2)
if (require("qvalue")) {
  set.seed(2014)

  # generate p-values from many one sample t-tests: half of them null
  oracle <- rep(c(0, .5), each=1000)
  pvals <- sapply(oracle, function(mu) t.test(rnorm(15, mu))$p.value)
  qplot(pvals)

  q <- qvalue(pvals)

  tidy(q)
  head(augment(q))
  glance(q)

  # use augmented data to compare p-values to q-values
  ggplot(augment(q), aes(p.value, q.value)) + geom_point()

  # use tidy see how pi0 estimate changes with lambda, comparing
  # to smoothed version
  g <- ggplot(tidy(q), aes(lambda, pi0, color=smoothed)) + geom_line()
  g

  # show the chosen value
  g + geom_hline(yintercept=q$pi0, lty=2)
}
```

---

SummarizedExperiment\_tidiers

*Tidying methods for Biobase's SummarizedExperiment objects*


---

## Description

Tidying methods for Biobase's SummarizedExperiment objects

## Usage

```
## S3 method for class 'RangedSummarizedExperiment'
tidy(x, addPheno = FALSE,
     assay = SummarizedExperiment::assayNames(x)[1L], ...)
```

## Arguments

x	SummarizedExperiment object
addPheno	whether columns should be included in the tidied output for those in the SummarizedExperiment colData
assay	Which assay to return as the value column. Defaults to assays(x)[[1L]]
...	extra arguments (not used)

## Details

addPheno=TRUE adds columns that are redundant (since they add per-sample information to a per-sample-per-gene data frame), but that are useful for some kinds of graphs and analyses.

## Value

tidy returns a data frame with one row per gene-sample combination, with columns

gene	gene name
sample	sample name (from column names)
value	expressions

If addPheno is TRUE then information from colData is added.

## Examples

```
if (require("SummarizedExperiment", "airway")) {
  data(airway)

  se <- airway
  tidy(se)
}
```

---

tidy.deSet	<i>Tidying methods for edge's deSet object</i>
------------	------------------------------------------------

---

**Description**

Tidying methods for edge's deSet object

**Usage**

```
## S3 method for class 'deSet'
tidy(x, addPheno = FALSE, ...)
```

```
## S3 method for class 'deSet'
augment(x, data, ...)
```

```
## S3 method for class 'deSet'
glance(x, ...)
```

**Arguments**

x	deSet object
addPheno	whether columns should be included in the tidied output for those in the ExpressionSet's phenoData
...	extra arguments (not used)
data	Original data can be added. Default is NULL.

**Details**

addPheno=TRUE adds columns that are redundant (since they add per-sample information to a per-sample-per-gene data frame), but that are useful for some kinds of graphs and analyses.

**Value**

tidy returns a data frame with one row per gene-sample combination, with columns

gene	gene name
sample	sample name (from column names)
value	expressions on log2 scale

augment returns a data.frame with

p.value	the original p-values given to qvalue
q.value	the computed q-values
lfdr	the local false discovery rate

glance returns a data.frame with the model fits

---

`tidy.GRanges`*Tidying methods for GRanges and GRangesList objects.*

---

## Description

Tidying methods for GRanges and GRangesList objects.

## Usage

```
## S3 method for class 'GRanges'  
tidy(x, ...)  
  
## S3 method for class 'GRangesList'  
tidy(x, ...)  
  
## S3 method for class 'GRanges'  
glance(x, ...)  
  
## S3 method for class 'GRangesList'  
glance(x, ...)
```

## Arguments

<code>x</code>	GRanges or GRangesList object
<code>...</code>	Not used.

## Value

All tidying methods return a `data.frame` without rownames. `tidy` returns one row for each range, which contains

- start of the range
- end of the range
- width (or length) of the range
- names of the range
- strand
- seqname Name of the sequence from which the range comes (usually the chromosome)
- metadata Any included metadata, (ie, score, GC content)

For `GRangesList`, there will also be a column representing which group the ranges comes from. `glance` returns a `data.frame` with the number of ranges, the number of sequences, and the number of groups (if applicable).



**Examples**

```
if (require("GenomicRanges", "airway")) {
  data(airway)

  # GRangesList object
  air_gr <- rowRanges(airway)

  tidy(air_gr)
  glance(air_gr)

  # GRanges object
  air_gr <- rowRanges(airway)@unlistData

  tidy(air_gr)
  glance(air_gr)

}
```

# Index

- \* **datasets**
  - hammer, 7
- assayDataElement, 6
- augment.deSet (tidy.deSet), 15
- augment.DGEList (edgeR\_tidiers), 5
- augment.MArrayLM (limma\_tidiers), 8
- augment.qvalue (qvalue\_tidiers), 12
- augment\_sva, 2
  
- biobroom, 3
- biobroom-package (biobroom), 3
  
- DESeq2\_tidiers, 3
  
- edgeR\_tidiers, 5
- ExpressionSet\_tidiers, 6
  
- glance.deSet (tidy.deSet), 15
- glance.DGEEExact (edgeR\_tidiers), 5
- glance.GRanges (tidy.GRanges), 16
- glance.GRangesList (tidy.GRanges), 16
- glance.list (list\_tidiers), 10
- glance.MArrayLM (limma\_tidiers), 8
- glance.qvalue (qvalue\_tidiers), 12
- glance\_sva (augment\_sva), 2
  
- hammer, 7
  
- limma\_tidiers, 8
- list\_tidiers, 10
  
- MSnSet\_tidiers, 11
  
- qvalue\_tidiers, 12
  
- SummarizedExperiment\_tidiers, 14
  
- tidy.DESeqDataSet (DESeq2\_tidiers), 3
- tidy.DESeqResults (DESeq2\_tidiers), 3
- tidy.deSet, 15
- tidy.DGEEExact (edgeR\_tidiers), 5
- tidy.DGEList (edgeR\_tidiers), 5
- tidy.EList (limma\_tidiers), 8
- tidy.ExpressionSet (ExpressionSet\_tidiers), 6
- tidy.GRanges, 16
- tidy.GRangesList (tidy.GRanges), 16
- tidy.list (list\_tidiers), 10
- tidy.MAList (limma\_tidiers), 8
- tidy.MArrayLM (limma\_tidiers), 8
- tidy.MSnSet (MSnSet\_tidiers), 11
- tidy.qvalue (qvalue\_tidiers), 12
- tidy.RangedSummarizedExperiment (SummarizedExperiment\_tidiers), 14
- tidy\_sva (augment\_sva), 2