# Package 'tidySummarizedExperiment'

October 24, 2025

Type Package

Title Brings SummarizedExperiment to the Tidyverse

**Version** 1.19.7

Author Stefano Mangiola [aut, cre] <mangiolastefano@gmail.com>

Maintainer Stefano Mangiola <mangiolastefano@gmail.com>

**Description** The tidySummarizedExperiment package provides a set of tools for creating and manipulating tidy data representations of SummarizedExperiment objects. SummarizedExperiment

is a widely used data structure in bioinformatics for storing high-throughput genomic data, such as gene expression or DNA sequencing data.

The tidySummarizedExperiment package introduces a tidy framework for working with SummarizedExperiment objects.

It allows users to convert their data into a tidy format, where each observation is a row and each variable is a column. This tidy representation simplifies data manipulation, integration with other tidyverse packages, and enables seamless integration with the broader ecosystem of tidy tools for data analysis.

License GPL-3

**Depends** R (>= 4.3.0), SummarizedExperiment, ttservice (>= 0.5.0)

**Imports** dplyr, tibble (>= 3.0.4), magrittr, tidyr, ggplot2, rlang, purrr, lifecycle, methods, utils, S4Vectors, tidyselect, ellipsis, vctrs, pillar, stringr, cli, fansi, stats, pkgconfig, plyxp

**Suggests** BiocStyle, testthat, knitr, markdown, rmarkdown, plotly, rbibutils, prettydoc, airway

VignetteBuilder knitr

RdMacros lifecycle

Biarch true

**biocViews** AssayDomain, Infrastructure, RNASeq, DifferentialExpression, GeneExpression, Normalization, Clustering, QualityControl, Sequencing, Transcription, Transcriptomics

**Encoding** UTF-8

LazyData true

RoxygenNote 7.3.2

**Roxygen** list(markdown = TRUE)

2 Contents

# LazyDataCompression xz

 ${\bf URL}\ {\tt https://github.com/stemangiola/tidySummarizedExperiment}$ 

**BugReports** https://github.com/stemangiola/tidySummarizedExperiment/issues

 $\label{lem:conductor} \textbf{git\_url} \ \ \text{https://git.bioconductor.org/packages/tidySummarizedExperiment}$ 

git\_branch devel

git\_last\_commit 9f7c26e

git\_last\_commit\_date 2025-10-12

**Repository** Bioconductor 3.22 **Date/Publication** 2025-10-23

# **Contents**

Index

append_samples	
as_tibble	
oind_rows	
count	
listinct	
extract	
ilter	1
`ull_join	1
ggplot	1.
group_by	1
group_split	1
nner_join	1
eft_join	
nutate	2
nest	2
pasilla	2
pivot_longer	2
pivot_wider	3
olot_ly	3
pull	3
ename	3
ight_join	3
owwise	4
sample_n	4
se	4
select	4
separate	4
slice	5
summarise	5
idy	
inite	
ınnest	
%>%	5

**60** 

append\_samples 3

append\_samples

Append samples from multiple SummarizedExperiment objects

# Description

Append samples from multiple SummarizedExperiment objects by column-binding them. This function is equivalent to cbind but provides a tidyverse-like interface.

## Usage

```
## S3 method for class 'SummarizedExperiment'
append_samples(x, ..., .id = NULL)
```

#### **Arguments**

x First SummarizedExperiment object to combine

... Additional SummarizedExperiment objects to combine by samples

. id Object identifier (currently not used)

#### Value

A combined SummarizedExperiment object

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

#### **Examples**

```
data(se)
append_samples(se, se)
```

as\_tibble

Coerce lists, matrices, and more to data frames

### Description

as\_tibble() turns an existing object, such as a data frame or matrix, into a so-called tibble, a data frame with class tbl\_df. This is in contrast with tibble(), which builds a tibble from individual columns. as\_tibble() is to tibble() as base::as.data.frame() is to base::data.frame().

as\_tibble() is an S3 generic, with methods for:

- data.frame: Thin wrapper around the list method that implements tibble's treatment of rownames.
- matrix, poly, ts, table
- Default: Other inputs are first coerced with base::as.data.frame().

4 as\_tibble

as\_tibble\_row() converts a vector to a tibble with one row. If the input is a list, all elements must have size one.

as\_tibble\_col() converts a vector to a tibble with one column.

#### Usage

```
## S3 method for class 'SummarizedExperiment'
as_tibble(
    x,
    ...,
    .name_repair = c("check_unique", "unique", "universal", "minimal"),
    rownames = pkgconfig::get_config("tibble::rownames", NULL)
)
```

#### **Arguments**

x A data frame, list, matrix, or other object that could reasonably be coerced to a

... Unused, for extensibility.

.name\_repair

Treatment of problematic column names:

- "minimal": No name repair or checks, beyond basic existence,
- "unique": Make sure names are unique and not empty,
- "check\_unique": (default value), no name repair, but check they are unique,
- "universal": Make the names unique and syntactic
- "unique\_quiet": Same as "unique", but "quiet"
- "universal\_quiet": Same as "universal", but "quiet"
- a function: apply custom name repair (e.g., .name\_repair = make.names for names in the style of base R).
- A purrr-style anonymous function, see rlang::as\_function()

This argument is passed on as repair to vctrs::vec\_as\_names(). See there for more details on these terms and the strategies used to enforce them.

rownames

How to treat existing row names of a data frame or matrix:

- NULL: remove row names. This is the default.
- NA: keep row names.
- A string: the name of a new column. Existing rownames are transferred into this column and the row.names attribute is deleted. No name repair is applied to the new column name, even if x already contains a column of that name. Use as\_tibble(rownames\_to\_column(...)) to safeguard against this case.

Read more in rownames.

### Value

tibble

#### **Row names**

The default behavior is to silently remove row names.

New code should explicitly convert row names to a new column using the rownames argument.

For existing code that relies on the retention of row names, call pkgconfig::set\_config("tibble::rownames" = NA) in your script or in your package's .onLoad() function.

bind\_rows 5

### Life cycle

Using as\_tibble() for vectors is superseded as of version 3.0.0, prefer the more expressive as\_tibble\_row() and as\_tibble\_col() variants for new code.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Müller, K., Wickham, H. (2023). tibble: Simple Data Frames. R package version 3.2.1, https://CRAN.R-project.org/package=tibble

#### See Also

tibble() constructs a tibble from individual columns. enframe() converts a named vector to a tibble with a column of names and column of values. Name repair is implemented using vctrs::vec\_as\_names().

### **Examples**

```
tidySummarizedExperiment::pasilla %>%
    as_tibble()

tidySummarizedExperiment::pasilla %>%
    as_tibble(.subset=-c(condition, type))
```

bind\_rows

Efficiently bind multiple data frames by row and column

# Description

This is an efficient implementation of the common pattern of 'do.call(rbind, dfs)' or 'do.call(cbind, dfs)' for binding many data frames into one.

This is an efficient implementation of the common pattern of 'do.call(rbind, dfs)' or 'do.call(cbind, dfs)' for binding many data frames into one.

### Usage

```
## S3 method for class 'SummarizedExperiment'
bind_rows(..., .id = NULL, add.cell.ids = NULL)
## S3 method for class 'SummarizedExperiment'
bind_cols(..., .id = NULL)
## S3 method for class 'RangedSummarizedExperiment'
bind_cols(..., .id = NULL)
```

6 bind\_rows

### **Arguments**

... Data frames to combine.

Each argument can either be a data frame, a list that could be a data frame, or a list of data frames.

When row-binding, columns are matched by name, and any missing columns will be filled with NA.

When column-binding, rows are matched by position, so all data frames must have the same number of rows. To match by value, not position, see mutate-joins.

.id Data frame identifier.

When '.id' is supplied, a new column of identifiers is created to link each row to its original data frame. The labels are taken from the named arguments to 'bind\_rows()'. When a list of data frames is supplied, the labels are taken from the names of the list. If no names are found a numeric sequence is used instead.

add.cell.ids Appends the corresponding values to

#### **Details**

The output of 'bind\_rows()' will contain a column if that column appears in any of the inputs.

The output of 'bind\_rows()' will contain a column if that column appears in any of the inputs.

#### Value

'bind\_rows()' and 'bind\_cols()' return the same type as the first input, either a data frame, 'tbl\_df', or 'grouped\_df'.

'bind\_rows()' and 'bind\_cols()' return the same type as the first input, either a data frame, 'tbl\_df', or 'grouped\_df'.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

### **Examples**

```
print("small_pbmc |> bind_rows(small_pbmc)")
print("small_pbmc |> bind_cols(annotation_column)")
```

count 7

count

Count the observations in each group

### **Description**

count() lets you quickly count the unique values of one or more variables: df %>% count(a, b) is roughly equivalent to df %>% group\_by(a, b) %>% summarise(n = n()). count() is paired with tally(), a lower-level helper that is equivalent to df %>% summarise(n = n()). Supply wt to perform weighted counts, switching the summary from n = n() to n = sum(wt).

add\_count() and add\_tally() are equivalents to count() and tally() but use mutate() instead of summarise() so that they add a new column with group-wise counts.

## Usage

```
## $3 method for class 'SummarizedExperiment'
count(
    x,
    ...,
    wt = NULL,
    sort = FALSE,
    name = NULL,
    .drop = group_by_drop_default(x)
)
```

### **Arguments**

A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr).

... <data-masking> Variables to group by.

wt <data-masking> Frequency weights. Can be NULL or a variable:

• If NULL (the default), counts the number of rows in each group.

• If a variable, computes sum(wt) for each group.

sort If TRUE, will show the largest groups at the top.

name The name of the new column in the output.

If omitted, it will default to n. If there's already a column called n, it will use nn. If there's a column called n and nn, it'll use nnn, and so on, adding ns until

it gets a new name.

Handling of factor levels that don't appear in the data, passed on to group\_by().

For count(): if FALSE will include counts for empty groups (i.e. for levels of

factors that don't exist in the data).

[Deprecated] For add\_count(): deprecated since it can't actually affect the output.

#### Value

.drop

An object of the same type as .data. count() and add\_count() group transiently, so the output has the same groups as the input.

8 distinct

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

### **Examples**

```
data(se)
se |> count(dex)
```

distinct

Keep distinct/unique rows

### **Description**

Keep only unique/distinct rows from a data frame. This is similar to unique.data.frame() but considerably faster.

# Usage

```
## S3 method for class 'SummarizedExperiment'
distinct(.data, ..., .keep_all = FALSE)
```

### **Arguments**

.data	A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See <i>Methods</i> , below, for more details.
	<data-masking> Optional variables to use when determining uniqueness. If there are multiple rows for a given combination of inputs, only the first row will be preserved. If omitted, will use all variables in the data frame.</data-masking>
.keep_all	If TRUE, keep all variables in .data. If a combination of is not distinct, this keeps the first row of values.

### Value

An object of the same type as .data. The output has the following properties:

- Rows are a subset of the input but appear in the same order.
- Columns are not modified if . . . is empty or .keep\_all is TRUE. Otherwise, distinct() first calls mutate() to create new columns.
- Groups are not modified.
- Data frame attributes are preserved.

extract 9

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages: no methods found.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

# **Examples**

```
data(pasilla)
pasilla |> distinct(.sample)
```

extract

Extract a character column into multiple columns using regular expression groups

#### **Description**

### [Superseded]

extract() has been superseded in favour of separate\_wider\_regex() because it has a more polished API and better handling of problems. Superseded functions will not go away, but will only receive critical bug fixes.

Given a regular expression with capturing groups, extract() turns each group into a new column. If the groups don't match, or the input is NA, the output will be NA.

### Usage

```
## S3 method for class 'SummarizedExperiment'
extract(
  data,
  col,
  into,
  regex = "([[:alnum:]]+)",
  remove = TRUE,
  convert = FALSE,
  ...
)
```

10 filter

## **Arguments**

data	A data frame.
col	<tidy-select> Column to expand.</tidy-select>
into	Names of new variables to create as character vector. Use NA to omit the variable in the output. $$
regex	A string representing a regular expression used to extract the desired values. There should be one group (defined by ()) for each element of into.
remove	If TRUE, remove input column from output data frame.
convert	If TRUE, will run type.convert() with as.is = TRUE on new columns. This is useful if the component columns are integer, numeric or logical.  NB: this will cause string "NA"s to be converted to NAs.
	Additional arguments passed on to methods.

### Value

tidySummarizedExperiment

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

### See Also

```
separate() to split up by a separator.
```

### **Examples**

```
tidySummarizedExperiment::pasilla |>
    extract(type, into="sequencing", regex="([a-z]*)_end", convert=TRUE)
```

filter	Keep rows that match a condition	

# Description

The filter() function is used to subset a data frame, retaining all rows that satisfy your conditions. To be retained, the row must produce a value of TRUE for all conditions. Note that when a condition evaluates to NA the row will be dropped, unlike base subsetting with [.

### Usage

```
## S3 method for class 'SummarizedExperiment'
filter(.data, ..., .preserve = FALSE)
```

filter 11

#### **Arguments**

.data A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See Methods, below, for more details. <data-masking> Expressions that return a logical value, and are defined in . . . terms of the variables in .data. If multiple expressions are included, they are combined with the & operator. Only rows for which all conditions evaluate to TRUE are kept. Relevant when the .data input is grouped. If .preserve = FALSE (the default), .preserve the grouping structure is recalculated based on the resulting data, otherwise the

grouping is kept as is.

#### Details

The filter() function is used to subset the rows of .data, applying the expressions in ... to the column values to determine which rows should be retained. It can be applied to both grouped and ungrouped data (see group\_by() and ungroup()). However, dplyr is not yet smart enough to optimise the filtering operation on grouped datasets that do not need grouped calculations. For this reason, filtering is often considerably faster on ungrouped data.

#### Value

An object of the same type as .data. The output has the following properties:

- Rows are a subset of the input, but appear in the same order.
- · Columns are not modified.
- The number of groups may be reduced (if .preserve is not TRUE).
- Data frame attributes are preserved.

#### **Useful filter functions**

There are many functions and operators that are useful when constructing the expressions used to filter the data:

- ==, >, >= etc
- &, |, !, xor()
- is.na()
- between(), near()

# **Grouped tibbles**

Because filtering expressions are computed within groups, they may yield different results on grouped tibbles. This will be the case as soon as an aggregating, lagging, or ranking function is involved. Compare this ungrouped filtering:

```
starwars %>% filter(mass > mean(mass, na.rm = TRUE))
```

With the grouped equivalent:

```
starwars %>% group_by(gender) %>% filter(mass > mean(mass, na.rm = TRUE))
```

In the ungrouped version, filter() compares the value of mass in each row to the global average (taken over the whole data set), keeping only the rows with mass greater than this global average. In contrast, the grouped version calculates the average mass separately for each gender group, and keeps rows with mass greater than the relevant within-gender average.

12 full\_join

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages: no methods found.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

```
Other single table verbs: arrange(), mutate(), reframe(), rename(), select(), slice(), summarise()
```

# Examples

```
data(pasilla)
pasilla |> filter(.sample == "untrt1")
# Learn more in ?dplyr_tidy_eval
```

full\_join

Mutating joins

# Description

Mutating joins add columns from y to x, matching observations based on the keys. There are four mutating joins: the inner join, and the three outer joins.

#### Inner join:

An inner\_join() only keeps observations from x that have a matching key in y.

The most important property of an inner join is that unmatched rows in either input are not included in the result. This means that generally inner joins are not appropriate in most analyses, because it is too easy to lose observations.

# **Outer joins:**

The three outer joins keep observations that appear in at least one of the data frames:

- A left\_join() keeps all observations in x.
- A right\_join() keeps all observations in y.
- A full\_join() keeps all observations in x and y.

### Usage

```
## S3 method for class 'SummarizedExperiment'
full_join(x, y, by = NULL, copy = FALSE, suffix = c(".x", ".y"), ...)
```

full\_join 13

#### **Arguments**

by

x, y A pair of data frames, data frame extensions (e.g. a tibble), or lazy data frames (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

A join specification created with join\_by(), or a character vector of variables to join by.

If NULL, the default, \*\_join() will perform a natural join, using all variables in common across x and y. A message lists the variables so that you can check they're correct; suppress the message by supplying by explicitly.

To join on different variables between x and y, use a join\_by() specification. For example, join\_by(a == b) will match x\$a to y\$b.

To join by multiple variables, use a join\_by() specification with multiple expressions. For example, join\_by(a == b, c == d) will match x to y and x to y the column names are the same between x and y, you can shorten this by listing only the variable names, like join\_by(a, c).

join\_by() can also be used to perform inequality, rolling, and overlap joins.
See the documentation at ?join\_by for details on these types of joins.

For simple equality joins, you can alternatively specify a character vector of variable names to join by. For example, by = c("a", "b") joins x\$a to y\$a and x\$b to y\$b. If variable names differ between x and y, use a named character vector like by =  $c("x_a" = "y_a", "x_b" = "y_b")$ .

To perform a cross-join, generating all combinations of x and y, see cross\_join().

If x and y are not from the same data source, and copy is TRUE, then y will be copied into the same src as x. This allows you to join tables across srcs, but it is a potentially expensive operation so you must opt into it.

If there are non-joined duplicate variables in x and y, these suffixes will be added to the output to disambiguate them. Should be a character vector of length 2.

Other parameters passed onto methods.

### Value

An object of the same type as x (including the same groups). The order of the rows and columns of x is preserved as much as possible. The output has the following properties:

- The rows are affect by the join type.
  - inner\_join() returns matched x rows.
  - left\_join() returns all x rows.
  - right\_join() returns matched of x rows, followed by unmatched y rows.
  - full\_join() returns all x rows, followed by unmatched y rows.
- Output columns include all columns from x and all non-key columns from y. If keep = TRUE, the key columns from y are included as well.
- If non-key columns in x and y have the same name, suffixes are added to disambiguate. If keep = TRUE and key columns in x and y have the same name, suffixes are added to disambiguate these as well.
- If keep = FALSE, output columns included in by are coerced to their common type between x and y.

сору

suffix

14 full\_join

#### Many-to-many relationships

By default, dplyr guards against many-to-many relationships in equality joins by throwing a warning. These occur when both of the following are true:

- A row in x matches multiple rows in y.
- A row in y matches multiple rows in x.

This is typically surprising, as most joins involve a relationship of one-to-one, one-to-many, or many-to-one, and is often the result of an improperly specified join. Many-to-many relationships are particularly problematic because they can result in a Cartesian explosion of the number of rows returned from the join.

If a many-to-many relationship is expected, silence this warning by explicitly setting relationship = "many-to-many".

In production code, it is best to preemptively set relationship to whatever relationship you expect to exist between the keys of x and y, as this forces an error to occur immediately if the data doesn't align with your expectations.

Inequality joins typically result in many-to-many relationships by nature, so they don't warn on them by default, but you should still take extra care when specifying an inequality join, because they also have the capability to return a large number of rows.

Rolling joins don't warn on many-to-many relationships either, but many rolling joins follow a many-to-one relationship, so it is often useful to set relationship = "many-to-one" to enforce this.

Note that in SQL, most database providers won't let you specify a many-to-many relationship between two tables, instead requiring that you create a third *junction table* that results in two one-to-many relationships instead.

### Methods

These functions are **generics**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

- inner\_join(): no methods found.
- left\_join(): no methods found.
- right\_join(): no methods found.
- full\_join(): no methods found.

## References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

```
Other joins: cross_join(), filter-joins, nest_join()
```

ggplot 15

### **Examples**

```
data(pasilla)

tt <- pasilla

tt |> full_join(tibble::tibble(condition="treated", dose=10))
```

ggplot

Create a new ggplot from a SummarizedExperiment

#### **Description**

ggplot() initializes a ggplot object. It can be used to declare the input data frame for a graphic and to specify the set of plot aesthetics intended to be common throughout all subsequent layers unless specifically overridden.

### Usage

```
## S3 method for class 'SummarizedExperiment'
ggplot(data = NULL, mapping = aes(), ..., environment = parent.frame())
```

#### **Arguments**

Default dataset to use for plot. If not already a data frame, will be converted to
one by fortify(). If not specified, must be supplied in each layer added to the plot.
piot.
Default list of aesthetic mappings to use for plot. If not specified, must be sup-
plied in each layer added to the plot.
Other arguments passed on to methods. Not currently used.
[Deprecated] Used prior to tidy evaluation.

#### **Details**

ggplot() is used to construct the initial plot object, and is almost always followed by a plus sign (+) to add components to the plot.

There are three common patterns used to invoke ggplot():

```
    ggplot(data = df, mapping = aes(x, y, other aesthetics))
    ggplot(data = df)
    ggplot()
```

The first pattern is recommended if all layers use the same data and the same set of aesthetics, although this method can also be used when adding a layer using data from another data frame.

The second pattern specifies the default data frame to use for the plot, but no aesthetics are defined up front. This is useful when one data frame is used predominantly for the plot, but the aesthetics vary from one layer to another.

The third pattern initializes a skeleton ggplot object, which is fleshed out as layers are added. This is useful when multiple data frames are used to produce different layers, as is often the case in complex graphics.

The data = and mapping = specifications in the arguments are optional (and are often omitted in practice), so long as the data and the mapping values are passed into the function in the right order. In the examples below, however, they are left in place for clarity.

16 group\_by

#### Value

```
ggplot
```

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02200.2

Wickham, H. (2016). ggplot2: Elegant Graphics for Data Analysis. Springer-Verlag New York. ISBN 978-3-319-24277-4, https://ggplot2.tidyverse.org

#### See Also

The first steps chapter of the online ggplot2 book.

#### **Examples**

```
library(ggplot2)
data(pasilla)
pasilla %>%
    ggplot(aes(.sample, counts)) +
    geom_boxplot()
```

group\_by

Group by one or more variables

# Description

Most data operations are done on groups defined by variables. group\_by() takes an existing tbl and converts it into a grouped tbl where operations are performed "by group". ungroup() removes grouping.

# Usage

```
## S3 method for class 'SummarizedExperiment'
group_by(.data, ..., .add = FALSE, .drop = group_by_drop_default(.data))
```

### **Arguments**

.data

A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

. . .

In group\_by(), variables or computations to group by. Computations are always done on the ungrouped data frame. To perform computations on the grouped data, you need to use a separate mutate() step before the group\_by(). Computations are not allowed in nest\_by(). In ungroup(), variables to remove from the grouping.

.add

When FALSE, the default, group\_by() will override existing groups. To add to the existing groups, use .add = TRUE.

This argument was previously called add, but that prevented creating a new grouping variable called add, and conflicts with our naming conventions.

group\_by 17

.drop

Drop groups formed by factor levels that don't appear in the data? The default is TRUE except when .data has been previously grouped with .drop = FALSE. See group\_by\_drop\_default() for details.

#### Value

A grouped data frame with class grouped\_df, unless the combination of . . . and add yields a empty set of grouping columns, in which case a tibble will be returned.

### Methods

These function are **generics**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

```
• group_by(): no methods found.
```

• ungroup(): no methods found.

### **Ordering**

Currently, group\_by() internally orders the groups in ascending order. This results in ordered output from functions that aggregate groups, such as summarise().

When used as grouping columns, character vectors are ordered in the C locale for performance and reproducibility across R sessions. If the resulting ordering of your grouped operation matters and is dependent on the locale, you should follow up the grouped operation with an explicit call to arrange() and set the .locale argument. For example:

```
data %>%
  group_by(chr) %>%
  summarise(avg = mean(x)) %>%
  arrange(chr, .locale = "en")
```

This is often useful as a preliminary step before generating content intended for humans, such as an HTML table.

#### Legacy behavior:

Prior to dplyr 1.1.0, character vector grouping columns were ordered in the system locale. If you need to temporarily revert to this behavior, you can set the global option dplyr.legacy\_locale to TRUE, but this should be used sparingly and you should expect this option to be removed in a future version of dplyr. It is better to update existing code to explicitly call arrange(.locale = ) instead. Note that setting dplyr.legacy\_locale will also force calls to arrange() to use the system locale.

### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

18 group\_split

#### See Also

```
Other grouping functions: group_map(), group_nest(), group_split(), group_trim()
```

#### **Examples**

```
data(pasilla)
pasilla |> group_by(.sample)
```

group\_split

Split data frame by groups

# Description

### [Experimental]

group\_split() works like base::split() but:

- It uses the grouping structure from group\_by() and therefore is subject to the data mask
- It does not name the elements of the list based on the grouping as this only works well for a single character grouping variable. Instead, use group\_keys() to access a data frame that defines the groups.

group\_split() is primarily designed to work with grouped data frames. You can pass . . . to group and split an ungrouped data frame, but this is generally not very useful as you want have easy access to the group metadata.

#### Usage

```
## S3 method for class 'SummarizedExperiment'
group_split(.tbl, ..., .keep = TRUE)
```

### **Arguments**

```
.tbl A tbl.
... If .tbl is an ungrouped data frame, a grouping specification, forwarded to group_by().
.keep Should the grouping columns be kept?
```

### Value

A list of tibbles. Each tibble contains the rows of .tbl for the associated group and all the columns, including the grouping variables. Note that this returns a list\_of which is slightly stricter than a simple list but is useful for representing lists where every element has the same type.

# Lifecycle

group\_split() is not stable because you can achieve very similar results by manipulating the nested column returned from tidyr::nest(.by =). That also retains the group keys all within a single data structure. group\_split() may be deprecated in the future.

inner\_join 19

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

### See Also

```
Other grouping functions: group_by(), group_map(), group_nest(), group_trim()
```

### **Examples**

```
data(pasilla, package = "tidySummarizedExperiment")
pasilla |> group_split(condition)
pasilla |> group_split(counts > 0)
pasilla |> group_split(condition, counts > 0)
```

inner\_join

Mutating joins

### Description

Mutating joins add columns from y to x, matching observations based on the keys. There are four mutating joins: the inner join, and the three outer joins.

### Inner join:

An inner\_join() only keeps observations from x that have a matching key in y.

The most important property of an inner join is that unmatched rows in either input are not included in the result. This means that generally inner joins are not appropriate in most analyses, because it is too easy to lose observations.

### **Outer joins:**

The three outer joins keep observations that appear in at least one of the data frames:

- A left\_join() keeps all observations in x.
- A right\_join() keeps all observations in y.
- A full\_join() keeps all observations in x and y.

## Usage

```
## S3 method for class 'SummarizedExperiment'
inner_join(x, y, by = NULL, copy = FALSE, suffix = c(".x", ".y"), ...)
```

20 inner\_join

#### **Arguments**

by

x, y A pair of data frames, data frame extensions (e.g. a tibble), or lazy data frames (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

A join specification created with join\_by(), or a character vector of variables to join by.

If NULL, the default, \*\_join() will perform a natural join, using all variables in common across x and y. A message lists the variables so that you can check they're correct; suppress the message by supplying by explicitly.

To join on different variables between x and y, use a join\_by() specification. For example, join\_by(a == b) will match x\$a to y\$b.

To join by multiple variables, use a join\_by() specification with multiple expressions. For example, join\_by(a == b, c == d) will match x to y and x to y the column names are the same between x and y, you can shorten this by listing only the variable names, like join\_by(a, c).

join\_by() can also be used to perform inequality, rolling, and overlap joins.
See the documentation at ?join\_by for details on these types of joins.

For simple equality joins, you can alternatively specify a character vector of variable names to join by. For example, by = c("a", "b") joins x\$a to y\$a and x\$b to y\$b. If variable names differ between x and y, use a named character vector like by =  $c("x_a" = "y_a", "x_b" = "y_b")$ .

To perform a cross-join, generating all combinations of x and y, see cross\_join().

If x and y are not from the same data source, and copy is TRUE, then y will be copied into the same src as x. This allows you to join tables across srcs, but it is a potentially expensive operation so you must opt into it.

If there are non-joined duplicate variables in x and y, these suffixes will be added to the output to disambiguate them. Should be a character vector of length 2.

Other parameters passed onto methods.

### Value

An object of the same type as x (including the same groups). The order of the rows and columns of x is preserved as much as possible. The output has the following properties:

- The rows are affect by the join type.
  - inner\_join() returns matched x rows.
  - left\_join() returns all x rows.
  - right\_join() returns matched of x rows, followed by unmatched y rows.
  - full\_join() returns all x rows, followed by unmatched y rows.
- Output columns include all columns from x and all non-key columns from y. If keep = TRUE, the key columns from y are included as well.
- If non-key columns in x and y have the same name, suffixes are added to disambiguate. If keep = TRUE and key columns in x and y have the same name, suffixes are added to disambiguate these as well.
- If keep = FALSE, output columns included in by are coerced to their common type between x and y.

сору

suffix

inner\_join 21

### Many-to-many relationships

By default, dplyr guards against many-to-many relationships in equality joins by throwing a warning. These occur when both of the following are true:

- A row in x matches multiple rows in y.
- A row in y matches multiple rows in x.

This is typically surprising, as most joins involve a relationship of one-to-one, one-to-many, or many-to-one, and is often the result of an improperly specified join. Many-to-many relationships are particularly problematic because they can result in a Cartesian explosion of the number of rows returned from the join.

If a many-to-many relationship is expected, silence this warning by explicitly setting relationship = "many-to-many".

In production code, it is best to preemptively set relationship to whatever relationship you expect to exist between the keys of x and y, as this forces an error to occur immediately if the data doesn't align with your expectations.

Inequality joins typically result in many-to-many relationships by nature, so they don't warn on them by default, but you should still take extra care when specifying an inequality join, because they also have the capability to return a large number of rows.

Rolling joins don't warn on many-to-many relationships either, but many rolling joins follow a many-to-one relationship, so it is often useful to set relationship = "many-to-one" to enforce this.

Note that in SQL, most database providers won't let you specify a many-to-many relationship between two tables, instead requiring that you create a third *junction table* that results in two one-to-many relationships instead.

### Methods

These functions are **generics**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

- inner\_join(): no methods found.
- left\_join(): no methods found.
- right\_join(): no methods found.
- full\_join(): no methods found.

## References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

### See Also

```
Other joins: cross_join(), filter-joins, nest_join()
```

22 left\_join

#### **Examples**

```
data(pasilla)

tt <- pasilla

tt |> inner_join(tt |>
    distinct(condition) |>
    mutate(new_column=1:2) |>
    slice(1))
```

left\_join

Mutating joins

### Description

Mutating joins add columns from y to x, matching observations based on the keys. There are four mutating joins: the inner join, and the three outer joins.

#### Inner join:

An inner\_join() only keeps observations from x that have a matching key in y.

The most important property of an inner join is that unmatched rows in either input are not included in the result. This means that generally inner joins are not appropriate in most analyses, because it is too easy to lose observations.

### **Outer joins:**

The three outer joins keep observations that appear in at least one of the data frames:

- A left\_join() keeps all observations in x.
- A right\_join() keeps all observations in y.
- A full\_join() keeps all observations in x and y.

### Usage

```
## S3 method for class 'SummarizedExperiment'
left_join(x, y, by = NULL, copy = FALSE, suffix = c(".x", ".y"), ...)
```

### **Arguments**

x, y

A pair of data frames, data frame extensions (e.g. a tibble), or lazy data frames (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

by

A join specification created with join\_by(), or a character vector of variables to join by.

If NULL, the default, \*\_join() will perform a natural join, using all variables in common across x and y. A message lists the variables so that you can check they're correct; suppress the message by supplying by explicitly.

To join on different variables between x and y, use a  $join_by()$  specification. For example,  $join_by(a == b)$  will match x\$a to y\$b.

To join by multiple variables, use a  $join_by()$  specification with multiple expressions. For example,  $join_by(a == b, c == d)$  will match x\$a to y\$b and x\$c to y\$d. If the column names are the same between x and y, you can shorten this by listing only the variable names, like  $join_by(a, c)$ .

left\_join 23

join\_by() can also be used to perform inequality, rolling, and overlap joins.
See the documentation at ?join by for details on these types of joins.

For simple equality joins, you can alternatively specify a character vector of variable names to join by. For example, by = c("a", "b") joins x\$a to y\$a and x\$b to y\$b. If variable names differ between x and y, use a named character vector like by =  $c("x_a" = "y_a", "x_b" = "y_b")$ .

To perform a cross-join, generating all combinations of x and y, see cross\_join().

copy If x and y are not from the same data source, and copy is TRUE, then y will be copied into the same src as x. This allows you to join tables across srcs, but it is

a potentially expensive operation so you must opt into it.

suffix If there are non-joined duplicate variables in x and y, these suffixes will be added

to the output to disambiguate them. Should be a character vector of length 2.

.. Other parameters passed onto methods.

#### Value

An object of the same type as x (including the same groups). The order of the rows and columns of x is preserved as much as possible. The output has the following properties:

- The rows are affect by the join type.
  - inner\_join() returns matched x rows.
  - left\_join() returns all x rows.
  - right\_join() returns matched of x rows, followed by unmatched y rows.
  - full\_join() returns all x rows, followed by unmatched y rows.
- Output columns include all columns from x and all non-key columns from y. If keep = TRUE, the key columns from y are included as well.
- If non-key columns in x and y have the same name, suffixes are added to disambiguate. If keep = TRUE and key columns in x and y have the same name, suffixes are added to disambiguate these as well.
- If keep = FALSE, output columns included in by are coerced to their common type between x and y.

### Many-to-many relationships

By default, dplyr guards against many-to-many relationships in equality joins by throwing a warning. These occur when both of the following are true:

- A row in x matches multiple rows in y.
- A row in y matches multiple rows in x.

This is typically surprising, as most joins involve a relationship of one-to-one, one-to-many, or many-to-one, and is often the result of an improperly specified join. Many-to-many relationships are particularly problematic because they can result in a Cartesian explosion of the number of rows returned from the join.

If a many-to-many relationship is expected, silence this warning by explicitly setting relationship = "many-to-many".

In production code, it is best to preemptively set relationship to whatever relationship you expect to exist between the keys of x and y, as this forces an error to occur immediately if the data doesn't align with your expectations.

24 left\_join

Inequality joins typically result in many-to-many relationships by nature, so they don't warn on them by default, but you should still take extra care when specifying an inequality join, because they also have the capability to return a large number of rows.

Rolling joins don't warn on many-to-many relationships either, but many rolling joins follow a many-to-one relationship, so it is often useful to set relationship = "many-to-one" to enforce this.

Note that in SQL, most database providers won't let you specify a many-to-many relationship between two tables, instead requiring that you create a third *junction table* that results in two one-to-many relationships instead.

#### Methods

These functions are **generics**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

```
• inner_join(): no methods found.
```

- left\_join(): no methods found.
- right\_join(): no methods found.
- full\_join(): no methods found.

### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

```
Other joins: cross_join(), filter-joins, nest_join()
```

### **Examples**

```
data(pasilla)

tt <- pasilla

tt |> left_join(tt |>
    distinct(condition) |>
    mutate(new_column=1:2) |>
    slice(1))
```

mutate 25

mutate

Create, modify, and delete columns

#### **Description**

mutate() creates new columns that are functions of existing variables. It can also modify (if the name is the same as an existing column) and delete columns (by setting their value to NULL).

#### Usage

```
## S3 method for class 'SummarizedExperiment'
mutate(.data, ...)
```

### **Arguments**

.data

A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

... <data-masking> Name-value pairs. The name gives the name of the column in the output.

The value can be:

- A vector of length 1, which will be recycled to the correct length.
- A vector the same length as the current group (or the whole data frame if ungrouped).
- NULL, to remove the column.
- A data frame or tibble, to create multiple columns in the output.

#### Value

An object of the same type as .data. The output has the following properties:

- Columns from .data will be preserved according to the .keep argument.
- Existing columns that are modified by ... will always be returned in their original location.
- New columns created through . . . will be placed according to the . before and .after arguments.
- The number of rows is not affected.
- · Columns given the value NULL will be removed.
- Groups will be recomputed if a grouping variable is mutated.
- Data frame attributes are preserved.

### Useful mutate functions

- +, -, log(), etc., for their usual mathematical meanings
- lead(), lag()
- dense\_rank(), min\_rank(), percent\_rank(), row\_number(), cume\_dist(), ntile()
- cumsum(), cummean(), cummin(), cummax(), cumany(), cumall()
- na\_if(), coalesce()
- if\_else(), recode(), case\_when()

26 mutate

### **Grouped tibbles**

Because mutating expressions are computed within groups, they may yield different results on grouped tibbles. This will be the case as soon as an aggregating, lagging, or ranking function is involved. Compare this ungrouped mutate:

```
starwars %>%
  select(name, mass, species) %>%
  mutate(mass_norm = mass / mean(mass, na.rm = TRUE))

With the grouped equivalent:

starwars %>%
  select(name, mass, species) %>%
  group_by(species) %>%
  mutate(mass_norm = mass / mean(mass, na.rm = TRUE))
```

The former normalises mass by the global average whereas the latter normalises by the averages within species levels.

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages: no methods found.

### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

```
Other single table verbs: rename(), slice(), summarise()
```

### **Examples**

```
data(pasilla)
pasilla |> mutate(logcounts=log2(counts))
```

nest 27

nest

Nest rows into a list-column of data frames

### **Description**

Nesting creates a list-column of data frames; unnesting flattens it back out into regular columns. Nesting is implicitly a summarising operation: you get one row for each group defined by the nonnested columns. This is useful in conjunction with other summaries that work with whole datasets, most notably models.

Learn more in vignette("nest").

### Usage

```
## S3 method for class 'SummarizedExperiment'
nest(.data, ..., .names_sep = NULL)
```

#### **Arguments**

.data A data frame.

... <tidy-select> Columns to nest; these will appear in the inner data frames.

Specified using name-variable pairs of the form  $new_col = c(col1, col2, col3)$ .

The right hand side can be any valid tidyselect expression.

If not supplied, then  $\dots$  is derived as all columns not selected by .by, and will

use the column name from .key.

[Deprecated]: previously you could write df %>% nest(x, y, z). Convert to

df % nest(data = c(x, y, z)).

.names\_sep If NULL, the default, the inner names will come from the former outer names. If

a string, the new inner names will use the outer names with names\_sep automatically stripped. This makes names\_sep roughly symmetric between nesting

and unnesting.

### **Details**

If neither . . . nor .by are supplied, nest() will nest all variables, and will use the column name supplied through .key.

#### Value

tidySummarizedExperiment\_nested

#### New syntax

tidyr 1.0.0 introduced a new syntax for nest() and unnest() that's designed to be more similar to other functions. Converting to the new syntax should be straightforward (guided by the message you'll receive) but if you just need to run an old analysis, you can easily revert to the previous behaviour using nest\_legacy() and unnest\_legacy() as follows:

```
library(tidyr)
nest <- nest_legacy
unnest <- unnest_legacy</pre>
```

28 pasilla

#### **Grouped data frames**

df %>% nest(data = c(x, y)) specifies the columns to be nested; i.e. the columns that will appear in the inner data frame. df %>% nest(.by = c(x, y)) specifies the columns to nest by; i.e. the columns that will remain in the outer data frame. An alternative way to achieve the latter is to nest() a grouped data frame created by  $dplyr::group_by()$ . The grouping variables remain in the outer data frame and the others are nested. The result preserves the grouping of the input.

Variables supplied to nest() will override grouping variables so that df %% group\_by(x, y) %>% nest(data = !z) will be equivalent to df %% nest(data = !z).

You can't supply .by with a grouped data frame, as the groups already represent what you are nesting by.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

### **Examples**

```
tidySummarizedExperiment::pasilla |>
   nest(data=-condition)
```

pasilla

Read counts of RNA-seq samples of Pasilla knock-down by Brooks et al.

#### **Description**

A SummarizedExperiment dataset containing the transcriptome information for Drosophila Melanogaster.

### Usage

```
data(pasilla)
```

#### **Format**

containing 14599 features and 7 biological replicates.

### Source

https://bioconductor.org/packages/release/data/experiment/html/pasilla.html

pivot\_longer 29

pivot\_longer

Pivot data from wide to long

## **Description**

pivot\_longer() "lengthens" data, increasing the number of rows and decreasing the number of columns. The inverse transformation is pivot\_wider()

Learn more in vignette("pivot").

#### Usage

```
## S3 method for class 'SummarizedExperiment'
pivot_longer(
  data,
  cols.
  cols_vary = "fastest",
  names_to = "name",
  names_prefix = NULL,
  names_sep = NULL,
  names_pattern = NULL,
  names_ptypes = NULL,
  names_transform = NULL,
  names_repair = "check_unique",
  values_to = "value",
  values_drop_na = FALSE,
  values_ptypes = NULL,
  values_transform = NULL
```

### **Arguments**

data A data frame to pivot.

cols <tidy-select> Columns to pivot into longer format.

... Additional arguments passed on to methods.

raditional arguments passed on to methods.

When pivoting cols into longer format, how should the output rows be arranged relative to their original row number?

- "fastest", the default, keeps individual rows from cols close together in the output. This often produces intuitively ordered output when you have at least one key column from data that is not involved in the pivoting process.
- "slowest" keeps individual columns from cols close together in the output. This often produces intuitively ordered output when you utilize all of the columns from data in the pivoting process.

names\_to

cols\_vary

A character vector specifying the new column or columns to create from the information stored in the column names of data specified by cols.

- If length 0, or if NULL is supplied, no columns will be created.
- If length 1, a single column will be created which will contain the column names specified by cols.

30 pivot\_longer

• If length > 1, multiple columns will be created. In this case, one of names\_sep or names\_pattern must be supplied to specify how the column names should be split. There are also two additional character values you can take advantage of:

- NA will discard the corresponding component of the column name.
- ".value" indicates that the corresponding component of the column name defines the name of the output column containing the cell values, overriding values\_to entirely.

names\_prefix A regular expression used to remove matching text from the start of each variable name.

 ${\tt names\_sep, names\_pattern}$ 

If names\_to contains multiple values, these arguments control how the column name is broken up.

names\_sep takes the same specification as separate(), and can either be a numeric vector (specifying positions to break on), or a single string (specifying a regular expression to split on).

names\_pattern takes the same specification as extract(), a regular expression containing matching groups (()).

If these arguments do not give you enough control, use pivot\_longer\_spec() to create a spec object and process manually as needed.

### names\_ptypes, values\_ptypes

Optionally, a list of column name-prototype pairs. Alternatively, a single empty prototype can be supplied, which will be applied to all columns. A prototype (or ptype for short) is a zero-length vector (like integer() or numeric()) that defines the type, class, and attributes of a vector. Use these arguments if you want to confirm that the created columns are the types that you expect. Note that if you want to change (instead of confirm) the types of specific columns, you should use names\_transform or values\_transform instead.

#### names\_transform, values\_transform

Optionally, a list of column name-function pairs. Alternatively, a single function can be supplied, which will be applied to all columns. Use these arguments if you need to change the types of specific columns. For example, names\_transform = list(week = as.integer) would convert a character variable called week to an integer.

If not specified, the type of the columns generated from names\_to will be character, and the type of the variables generated from values\_to will be the common type of the input columns used to generate them.

names\_repair

What happens if the output has invalid column names? The default, "check\_unique" is to error if the columns are duplicated. Use "minimal" to allow duplicates in the output, or "unique" to de-duplicated by adding numeric suffixes. See vctrs::vec\_as\_names() for more options.

values\_to

A string specifying the name of the column to create from the data stored in cell values. If names\_to is a character containing the special .value sentinel, this value will be ignored, and the name of the value column will be derived from part of the existing column names.

values\_drop\_na If TRUE, will drop rows that contain only NAs in the value\_to column. This effectively converts explicit missing values to implicit missing values, and should generally be used only when missing values in data were created by its structure.

pivot\_wider 31

#### **Details**

pivot\_longer() is an updated approach to gather(), designed to be both simpler to use and to handle more use cases. We recommend you use pivot\_longer() for new code; gather() isn't going away but is no longer under active development.

#### Value

tidySummarizedExperiment

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

### **Examples**

```
# See vignette("pivot") for examples and explanation
library(dplyr)
tidySummarizedExperiment::pasilla %>%
    pivot_longer(c(condition, type),
        names_to="name", values_to="value")
```

pivot\_wider

Pivot data from long to wide

# Description

pivot\_wider() "widens" data, increasing the number of columns and decreasing the number of rows. The inverse transformation is pivot\_longer().

Learn more in vignette("pivot").

### Usage

32 pivot\_wider

```
values_from = value,
  values fill = NULL.
  values_fn = NULL,
  unused_fn = NULL
)
```

### **Arguments**

data A data frame to pivot.

Additional arguments passed on to methods.

<tidy-select> A set of columns that uniquely identify each observation. Typid\_cols ically used when you have redundant variables, i.e. variables whose values are

perfectly correlated with existing variables.

Defaults to all columns in data except for the columns specified through names\_from and values\_from. If a tidyselect expression is supplied, it will be evaluated on data after removing the columns specified through names\_from and values\_from.

id\_expand Should the values in the id\_cols columns be expanded by expand() before pivoting? This results in more rows, the output will contain a complete expansion

of all possible values in id\_cols. Implicit factor levels that aren't represented in the data will become explicit. Additionally, the row values corresponding to

the expanded id\_cols will be sorted.

<tidy-select> A pair of arguments describing which column (or columns) to get the name of the output column (names\_from), and which column (or columns) to get the cell values from (values\_from).

If values\_from contains multiple values, the value will be added to the front of the output column.

names\_prefix String added to the start of every variable name. This is particularly useful

if names\_from is a numeric vector and you want to create syntactic variable

names.

If names\_from or values\_from contains multiple variables, this will be used to names\_sep

join their values together into a single string to use as a column name.

names\_glue Instead of names\_sep and names\_prefix, you can supply a glue specification that uses the names\_from columns (and special .value) to create custom col-

umn names.

Should the column names be sorted? If FALSE, the default, column names are names\_sort

ordered by first appearance.

When names\_from identifies a column (or columns) with multiple unique valnames\_vary ues, and multiple values\_from columns are provided, in what order should the

resulting column names be combined?

• "fastest" varies names\_from values fastest, resulting in a column naming scheme of the form: value1\_name1, value1\_name2, value2\_name1, value2\_name2. This is the default.

• "slowest" varies names\_from values slowest, resulting in a column naming scheme of the form: value1\_name1, value2\_name1, value1\_name2, value2\_name2.

names\_expand

Should the values in the names\_from columns be expanded by expand() before pivoting? This results in more columns, the output will contain column names corresponding to a complete expansion of all possible values in names\_from. Implicit factor levels that aren't represented in the data will become explicit.

names\_from, values\_from

pivot\_wider 33

Additionally, the column names will be sorted, identical to what names\_sort would produce.

names\_repair

What happens if the output has invalid column names? The default, "check\_unique" is to error if the columns are duplicated. Use "minimal" to allow duplicates in the output, or "unique" to de-duplicated by adding numeric suffixes. See vctrs::vec\_as\_names() for more options.

values\_fill

Optionally, a (scalar) value that specifies what each value should be filled in with when missing.

This can be a named list if you want to apply different fill values to different value columns.

values\_fn

Optionally, a function applied to the value in each cell in the output. You will typically use this when the combination of id\_cols and names\_from columns does not uniquely identify an observation.

This can be a named list if you want to apply different aggregations to different values\_from columns.

unused\_fn

Optionally, a function applied to summarize the values from the unused columns (i.e. columns not identified by id\_cols, names\_from, or values\_from).

The default drops all unused columns from the result.

This can be a named list if you want to apply different aggregations to different unused columns.

id\_cols must be supplied for unused\_fn to be useful, since otherwise all unspecified columns will be considered id\_cols.

This is similar to grouping by the id\_cols then summarizing the unused columns using unused\_fn.

#### **Details**

pivot\_wider() is an updated approach to spread(), designed to be both simpler to use and to handle more use cases. We recommend you use pivot\_wider() for new code; spread() isn't going away but is no longer under active development.

#### Value

 ${\tt tidySummarizedExperiment}$ 

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

#### See Also

pivot\_wider\_spec() to pivot "by hand" with a data frame that defines a pivoting specification.

34 plot\_ly

#### **Examples**

```
# See vignette("pivot") for examples and explanation
library(dplyr)
tidySummarizedExperiment::pasilla %>%
    pivot_wider(names_from=feature, values_from=counts)
```

plot\_ly

Initiate a plotly visualization

#### **Description**

This function maps R objects to plotly.js, an (MIT licensed) web-based interactive charting library. It provides abstractions for doing common things (e.g. mapping data values to fill colors (via color) or creating animations (via frame)) and sets some different defaults to make the interface feel more 'R-like' (i.e., closer to plot() and ggplot2::qplot()).

### Usage

```
## S3 method for class 'tbl_df'
plot_ly(
  data = data.frame(),
  type = NULL,
  name = NULL,
  color = NULL,
  colors = NULL,
  alpha = NULL,
  stroke = NULL,
  strokes = NULL,
  alpha_stroke = 1,
  size = NULL,
  sizes = c(10, 100),
  span = NULL,
  spans = c(1, 20),
  symbol = NULL,
  symbols = NULL,
  linetype = NULL,
  linetypes = NULL,
  split = NULL,
  frame = NULL,
  width = NULL,
  height = NULL,
  source = "A"
)
## S3 method for class 'SummarizedExperiment'
plot_ly(
  data = data.frame(),
  type = NULL,
```

plot\_ly 35

```
name = NULL,
  color = NULL.
  colors = NULL,
  alpha = NULL,
  stroke = NULL,
  strokes = NULL,
  alpha_stroke = 1,
  size = NULL,
  sizes = c(10, 100),
  span = NULL,
  spans = c(1, 20),
  symbol = NULL,
  symbols = NULL,
  linetype = NULL,
  linetypes = NULL,
  split = NULL,
  frame = NULL,
 width = NULL,
 height = NULL,
  source = "A"
)
```

#### **Arguments**

data A data frame (optional) or crosstalk::SharedData object.

Arguments (i.e., attributes) passed along to the trace type. See schema() for a list of acceptable attributes for a given trace type (by going to traces -> type -> attributes). Note that attributes provided at this level may override other arguments (e.g. plot\_ly(x = 1:10, y = 1:10, color = I("red"),

marker = list(color = "blue"))).

type A character string specifying the trace type (e.g. "scatter", "bar", "box",

etc). If specified, it always creates a trace, otherwise

name Values mapped to the trace's name attribute. Since a trace can only have one

name, this argument acts very much like split in that it creates one trace for

every unique value.

color Values mapped to relevant 'fill-color' attribute(s) (e.g. fillcolor, marker.color,

textfont.color, etc.). The mapping from data values to color codes may be controlled using colors and alpha, or avoided altogether via I() (e.g., color = I("red")). Any color understood by grDevices::col2rgb() may be used in

this way.

colors Either a colorbrewer2.org palette name (e.g. "YlOrRd" or "Blues"), or a vector

of colors to interpolate in hexadecimal "#RRGGBB" format, or a color interpo-

lation function like colorRamp().

alpha A number between 0 and 1 specifying the alpha channel applied to color. De-

faults to 0.5 when mapping to fillcolor and 1 otherwise.

stroke Similar to color, but values are mapped to relevant 'stroke-color' attribute(s)

(e.g., marker.line.color and line.color for filled polygons). If not specified, stroke

inherits from color.

strokes Similar to colors, but controls the stroke mapping.

alpha\_stroke Similar to alpha, but applied to stroke.

36 plot\_ly

size	(Numeric) values mapped to relevant 'fill-size' attribute(s) (e.g., marker.size, textfont.size, and error_x.width). The mapping from data values to symbols may be controlled using sizes, or avoided altogether via I() (e.g., size = I(30)).
sizes	A numeric vector of length 2 used to scale size to pixels.
span	(Numeric) values mapped to relevant 'stroke-size' attribute(s) (e.g., marker.line.width, line.width for filled polygons, and error_x.thickness) The mapping from data values to symbols may be controlled using spans, or avoided altogether via I() (e.g., span = I(30)).
spans	A numeric vector of length 2 used to scale span to pixels.
symbol	(Discrete) values mapped to marker.symbol. The mapping from data values to symbols may be controlled using symbols, or avoided altogether via I() (e.g., symbol = I("pentagon")). Any pch value or symbol name may be used in this way.
symbols	A character vector of pch values or symbol names.
linetype	(Discrete) values mapped to line.dash. The mapping from data values to symbols may be controlled using linetypes, or avoided altogether via I() (e.g., linetype = I("dash")). Any lty (see par) value or dash name may be used in this way.
linetypes	A character vector of 1ty values or dash names
split	(Discrete) values used to create multiple traces (one trace per value).
frame	(Discrete) values used to create animation frames.
width	Width in pixels (optional, defaults to automatic sizing).
height	Height in pixels (optional, defaults to automatic sizing).
source	a character string of length 1. Match the value of this string with the source argument in event_data() to retrieve the event data corresponding to a specific plot (shiny apps can have multiple plots).

### **Details**

Unless type is specified, this function just initiates a plotly object with 'global' attributes that are passed onto downstream uses of add\_trace() (or similar). A formula must always be used when referencing column name(s) in data (e.g. plot\_ly(mtcars,  $x = \sim wt$ )). Formulas are optional when supplying values directly, but they do help inform default axis/scale titles (e.g., plot\_ly(x = mtcarswt)) vs plot\_ly(x = mtcarswt))

### Value

plotly
plotly

### Author(s)

Carson Sievert

# References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

pull 37

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

#### See Also

- For initializing a plotly-geo object: plot\_geo()
- For initializing a plotly-mapbox object: plot\_mapbox()
- For translating a ggplot2 object to a plotly object: ggplotly()
- For modifying any plotly object: layout(), add\_trace(), style()
- For linked brushing: highlight()
- For arranging multiple plots: subplot(), crosstalk::bscols()
- For inspecting plotly objects: plotly\_json()
- For quick, accurate, and searchable plotly.js reference: schema()

## **Examples**

```
data(se)
se |>
    plot_ly(x = ~counts)

data(se)
se |>
    plot_ly(x = ~counts)
```

pull

Extract a single column

## **Description**

pull() is similar to \$. It's mostly useful because it looks a little nicer in pipes, it also works with remote data frames, and it can optionally name the output.

## Usage

```
## S3 method for class 'SummarizedExperiment'
pull(.data, var = -1, name = NULL, ...)
```

## **Arguments**

.data

A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

var

A variable specified as:

- a literal variable name
- a positive integer, giving the position counting from the left
- a negative integer, giving the position counting from the right.

38 rename

The default returns the last column (on the assumption that's the column you've created most recently).

This argument is taken by expression and supports quasiquotation (you can unquote column names and column locations).

name An optional parameter that specifies the column to be used as names for a named

vector. Specified in a similar manner as var.

... For use by methods.

#### Value

A vector the same size as .data.

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages: no methods found.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

## **Examples**

```
data(pasilla)
pasilla |> pull(feature)
```

rename

Rename columns

#### **Description**

rename() changes the names of individual variables using new\_name = old\_name syntax; rename\_with() renames columns using a function.

## Usage

```
## S3 method for class 'SummarizedExperiment' rename(.data, ...)
```

## Arguments

. . .

. data A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

For rename(): <tidy-select> Use new\_name = old\_name to rename selected variables.

For rename\_with(): additional arguments passed onto .fn.

right\_join 39

#### Value

An object of the same type as .data. The output has the following properties:

- · Rows are not affected.
- Column names are changed; column order is preserved.
- Data frame attributes are preserved.
- Groups are updated to reflect new names.

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages: no methods found.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

Other single table verbs: mutate(), slice(), summarise()

## **Examples**

```
data(pasilla)
pasilla |> rename(cond=condition)
```

right\_join

Mutating joins

#### **Description**

Mutating joins add columns from y to x, matching observations based on the keys. There are four mutating joins: the inner join, and the three outer joins.

#### Inner join:

An inner\_join() only keeps observations from x that have a matching key in y.

The most important property of an inner join is that unmatched rows in either input are not included in the result. This means that generally inner joins are not appropriate in most analyses, because it is too easy to lose observations.

## **Outer joins:**

The three outer joins keep observations that appear in at least one of the data frames:

- A left\_join() keeps all observations in x.
- A right\_join() keeps all observations in y.
- A full\_join() keeps all observations in x and y.

40 right\_join

#### Usage

```
## S3 method for class 'SummarizedExperiment' right_join(x, y, by = NULL, copy = FALSE, suffix = c(".x", ".y"), ...)
```

#### **Arguments**

x, y

A pair of data frames, data frame extensions (e.g. a tibble), or lazy data frames (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

by

A join specification created with join\_by(), or a character vector of variables to join by.

If NULL, the default, \*\_join() will perform a natural join, using all variables in common across x and y. A message lists the variables so that you can check they're correct; suppress the message by supplying by explicitly.

To join on different variables between x and y, use a  $join_by()$  specification. For example,  $join_by(a == b)$  will match x\$a to y\$b.

To join by multiple variables, use a join\_by() specification with multiple expressions. For example, join\_by(a == b, c == d) will match x to y and x to y the column names are the same between x and y, you can shorten this by listing only the variable names, like join\_by(a, c).

join\_by() can also be used to perform inequality, rolling, and overlap joins.
See the documentation at ?join\_by for details on these types of joins.

For simple equality joins, you can alternatively specify a character vector of variable names to join by. For example, by = c("a", "b") joins x\$a to y\$a and x\$b to y\$b. If variable names differ between x and y, use a named character vector like by =  $c("x_a" = "y_a", "x_b" = "y_b")$ .

To perform a cross-join, generating all combinations of x and y, see cross\_join().

copy

If x and y are not from the same data source, and copy is TRUE, then y will be copied into the same src as x. This allows you to join tables across srcs, but it is a potentially expensive operation so you must opt into it.

suffix

If there are non-joined duplicate variables in x and y, these suffixes will be added to the output to disambiguate them. Should be a character vector of length 2.

... Other parameters passed onto methods.

#### Value

An object of the same type as x (including the same groups). The order of the rows and columns of x is preserved as much as possible. The output has the following properties:

- The rows are affect by the join type.
  - inner\_join() returns matched x rows.
  - left\_join() returns all x rows.
  - right\_join() returns matched of x rows, followed by unmatched y rows.
  - full\_join() returns all x rows, followed by unmatched y rows.
- Output columns include all columns from x and all non-key columns from y. If keep = TRUE, the key columns from y are included as well.
- If non-key columns in x and y have the same name, suffixes are added to disambiguate. If keep = TRUE and key columns in x and y have the same name, suffixes are added to disambiguate these as well.
- If keep = FALSE, output columns included in by are coerced to their common type between x and y.

right\_join 41

#### Many-to-many relationships

By default, dplyr guards against many-to-many relationships in equality joins by throwing a warning. These occur when both of the following are true:

- A row in x matches multiple rows in y.
- A row in y matches multiple rows in x.

This is typically surprising, as most joins involve a relationship of one-to-one, one-to-many, or many-to-one, and is often the result of an improperly specified join. Many-to-many relationships are particularly problematic because they can result in a Cartesian explosion of the number of rows returned from the join.

If a many-to-many relationship is expected, silence this warning by explicitly setting relationship = "many-to-many".

In production code, it is best to preemptively set relationship to whatever relationship you expect to exist between the keys of x and y, as this forces an error to occur immediately if the data doesn't align with your expectations.

Inequality joins typically result in many-to-many relationships by nature, so they don't warn on them by default, but you should still take extra care when specifying an inequality join, because they also have the capability to return a large number of rows.

Rolling joins don't warn on many-to-many relationships either, but many rolling joins follow a many-to-one relationship, so it is often useful to set relationship = "many-to-one" to enforce this.

Note that in SQL, most database providers won't let you specify a many-to-many relationship between two tables, instead requiring that you create a third *junction table* that results in two one-to-many relationships instead.

## Methods

These functions are **generics**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

- inner\_join(): no methods found.
- left\_join(): no methods found.
- right\_join(): no methods found.
- full\_join(): no methods found.

## References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

## See Also

```
Other joins: cross_join(), filter-joins, nest_join()
```

42 rowwise

#### **Examples**

```
data(pasilla)

tt <- pasilla

tt |> right_join(tt |>
    distinct(condition) |>
    mutate(new_column=1:2) |>
    slice(1))
```

rowwise

Group input by rows

#### **Description**

rowwise() allows you to compute on a data frame a row-at-a-time. This is most useful when a vectorised function doesn't exist.

Most dplyr verbs preserve row-wise grouping. The exception is summarise(), which return a grouped\_df. You can explicitly ungroup with ungroup() or as\_tibble(), or convert to a grouped\_df with group\_by().

## Usage

```
## S3 method for class 'SummarizedExperiment'
rowwise(data, ...)
```

## **Arguments**

data Input data frame.

... <tidy-select> Variables to be preserved when calling summarise(). This is

typically a set of variables whose combination uniquely identify each row.

**NB**: unlike group\_by() you can not create new variables here but instead you can select multiple variables with (e.g.) everything().

## Value

A row-wise data frame with class rowwise\_df. Note that a rowwise\_df is implicitly grouped by row, but is not a grouped\_df.

#### **List-columns**

Because a rowwise has exactly one row per group it offers a small convenience for working with list-columns. Normally, summarise() and mutate() extract a groups worth of data with [. But when you index a list in this way, you get back another list. When you're working with a rowwise tibble, then dplyr will use [[ instead of [ to make your life a little easier.

## References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

sample\_n 43

#### See Also

nest\_by() for a convenient way of creating rowwise data frames with nested data.

#### **Examples**

# TODO

sample\_n

Sample n rows from a table

## **Description**

[Superseded] sample\_n() and sample\_frac() have been superseded in favour of slice\_sample(). While they will not be deprecated in the near future, retirement means that we will only perform critical bug fixes, so we recommend moving to the newer alternative.

These functions were superseded because we realised it was more convenient to have two mutually exclusive arguments to one function, rather than two separate functions. This also made it to clean up a few other smaller design issues with sample\_n()/sample\_frac:

- The connection to slice() was not obvious.
- The name of the first argument, tbl, is inconsistent with other single table verbs which use .data.
- The size argument uses tidy evaluation, which is surprising and undocumented.
- It was easier to remove the deprecated .env argument.
- ... was in a suboptimal position.

## Usage

```
## S3 method for class 'SummarizedExperiment'
sample_n(tbl, size, replace = FALSE, weight = NULL, .env = NULL, ...)
## S3 method for class 'SummarizedExperiment'
sample_frac(tbl, size = 1, replace = FALSE, weight = NULL, .env = NULL, ...)
```

## **Arguments**

tbl	A data.frame.	
size	<tidy-select> For sample_n(), the number of rows to select. For sample_frac(), the fraction of rows to select. If tbl is grouped, size applies to each group.</tidy-select>	
replace	Sample with or without replacement?	
weight	<tidy-select> Sampling weights. This must evaluate to a vector of non-negative numbers the same length as the input. Weights are automatically standardised to sum to 1.</tidy-select>	
.env	DEPRECATED.	
	ignored	

44 se

#### Value

tidySummarizedExperiment

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

## **Examples**

```
data(pasilla)
pasilla |> sample_n(50)
pasilla |> sample_frac(0.1)
```

se

Read counts of RNA-seq samples derived from Pasilla knock-down by Brooks et al.

#### **Description**

A SummarizedExperiment dataset containing the transcriptome information for Drosophila Melanogaster.

#### Usage

```
data(se)
```

#### **Format**

containing 14599 features and 7 biological replicates.

#### **Source**

https://bioconductor.org/packages/release/data/experiment/html/pasilla.html

select

Keep or drop columns using their names and types

#### **Description**

Select (and optionally rename) variables in a data frame, using a concise mini-language that makes it easy to refer to variables based on their name (e.g. a:f selects all columns from a on the left to f on the right) or type (e.g. where (is.numeric) selects all numeric columns).

#### Overview of selection features:

Tidyverse selections implement a dialect of R where operators make it easy to select variables:

- : for selecting a range of consecutive variables.
- ! for taking the complement of a set of variables.
- & and | for selecting the intersection or the union of two sets of variables.
- c() for combining selections.

In addition, you can use selection helpers. Some helpers select specific columns:

- everything(): Matches all variables.
- last\_col(): Select last variable, possibly with an offset.
- group\_cols(): Select all grouping columns.

Other helpers select variables by matching patterns in their names:

- starts\_with(): Starts with a prefix.
- ends\_with(): Ends with a suffix.
- contains(): Contains a literal string.
- matches(): Matches a regular expression.
- num\_range(): Matches a numerical range like x01, x02, x03.

Or from variables stored in a character vector:

- all\_of(): Matches variable names in a character vector. All names must be present, otherwise an out-of-bounds error is thrown.
- any\_of(): Same as all\_of(), except that no error is thrown for names that don't exist.

Or using a predicate function:

• where(): Applies a function to all variables and selects those for which the function returns TRUE.

## Usage

```
## S3 method for class 'SummarizedExperiment'
select(.data, ...)
```

## **Arguments**

. data A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

<tidy-select> One or more unquoted expressions separated by commas. Variable names can be used as if they were positions in the data frame, so expressions like x:y can be used to select a range of variables.

#### Value

An object of the same type as .data. The output has the following properties:

· Rows are not affected.

#> # i 146 more rows

- Output columns are a subset of input columns, potentially with a different order. Columns will be renamed if new\_name = old\_name form is used.
- Data frame attributes are preserved.
- Groups are maintained; you can't select off grouping variables.

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages: no methods found.

## **Examples**

Here we show the usage for the basic selection operators. See the specific help pages to learn about helpers like starts\_with().

The selection language can be used in functions like dplyr::select() or tidyr::pivot\_longer(). Let's first attach the tidyverse:

```
library(tidyverse)
# For better printing
iris <- as_tibble(iris)</pre>
Select variables by name:
starwars %>% select(height)
#> # A tibble: 87 x 1
#>
    height
#>
     <int>
#> 1
       172
#> 2
        167
#> 3
        96
#> 4
        202
#> # i 83 more rows
iris %>% pivot_longer(Sepal.Length)
#> # A tibble: 150 x 6
    Sepal.Width Petal.Length Petal.Width Species name
#>
                                                              value
#>
          <dbl> <dbl> <fct> <chr>
                                                              <dbl>
#> 1
            3.5
                         1.4
                                     0.2 setosa Sepal.Length
#> 2
            3
                         1.4
                                     0.2 setosa Sepal.Length
                                                                4.9
#> 3
            3.2
                         1.3
                                     0.2 setosa Sepal.Length
                                                                4.7
#> 4
            3.1
                         1.5
                                     0.2 setosa Sepal.Length
                                                                4.6
```

Select multiple variables by separating them with commas. Note how the order of columns is determined by the order of inputs:

```
starwars %>% select(homeworld, height, mass)
#> # A tibble: 87 x 3
#>
     homeworld height mass
                <int> <dbl>
#>
     <chr>
#> 1 Tatooine
                  172
                          77
#> 2 Tatooine
                  167
                          75
                          32
#> 3 Naboo
                   96
                  202
#> 4 Tatooine
                         136
#> # i 83 more rows
```

Functions like tidyr::pivot\_longer() don't take variables with dots. In this case use c() to select multiple variables:

```
iris %>% pivot_longer(c(Sepal.Length, Petal.Length))
#> # A tibble: 300 x 5
#>
     Sepal.Width Petal.Width Species name
                                                    value
                       <dbl> <fct>
                                                    <dbl>
#>
           <dbl>
                                      <chr>
#> 1
             3.5
                          0.2 setosa
                                      Sepal.Length
                                                      5.1
#> 2
             3.5
                          0.2 setosa
                                      Petal.Length
                                                      1.4
#> 3
             3
                          0.2 setosa
                                      Sepal.Length
                                                      4.9
#> 4
             3
                          0.2 setosa
                                      Petal.Length
                                                      1.4
#> # i 296 more rows
```

## **Operators::**

The: operator selects a range of consecutive variables:

```
starwars %>% select(name:mass)
#> # A tibble: 87 x 3
#>
    name
                    height mass
#>
     <chr>
                      <int> <dbl>
#> 1 Luke Skywalker
                        172
                               77
#> 2 C-3P0
                        167
                               75
#> 3 R2-D2
                         96
                               32
#> 4 Darth Vader
                        202
                              136
#> # i 83 more rows
```

The! operator negates a selection:

```
starwars %>% select(!(name:mass))
#> # A tibble: 87 x 11
#> hair_color skin_color eye_color birth_year sex gender
                                                              homeworld species
                                          <dbl> <chr> <chr>
#>
    <chr>
               <chr>
                           <chr>
                                                                <chr>
                                                                          <chr>
#> 1 blond
               fair
                           blue
                                          19
                                               male masculine Tatooine
                                                                          Human
#> 2 <NA>
               gold
                           yellow
                                          112
                                               none masculine Tatooine
                                                                          Droid
#> 3 <NA>
               white, blue red
                                           33
                                                none masculine Naboo
                                                                          Droid
#> 4 none
               white
                           yellow
                                          41.9 male masculine Tatooine Human
#> # i 83 more rows
#> # i 3 more variables: films <list>, vehicles <list>, starships <list>
iris %>% select(!c(Sepal.Length, Petal.Length))
#> # A tibble: 150 x 3
     Sepal.Width Petal.Width Species
           <dbl>
                       <dbl> <fct>
#>
#> 1
             3.5
                         0.2 setosa
```

```
#> 2
             3
                          0.2 setosa
#> 3
                          0.2 setosa
             3.2
#> 4
             3.1
                          0.2 setosa
#> # i 146 more rows
iris %>% select(!ends_with("Width"))
#> # A tibble: 150 x 3
     Sepal.Length Petal.Length Species
#>
            <dbl>
                          <dbl> <fct>
#> 1
              5.1
                            1.4 setosa
#> 2
              4.9
                            1.4 setosa
#> 3
              4.7
                            1.3 setosa
#> 4
              4.6
                            1.5 setosa
#> # i 146 more rows
& and | take the intersection or the union of two selections:
iris %>% select(starts_with("Petal") & ends_with("Width"))
#> # A tibble: 150 x 1
    Petal.Width
#>
           <dbl>
#>
#> 1
             0.2
#> 2
             0.2
#> 3
             0.2
#> 4
             0.2
#> # i 146 more rows
iris %>% select(starts_with("Petal") | ends_with("Width"))
#> # A tibble: 150 x 3
     Petal.Length Petal.Width Sepal.Width
#>
#>
            <dbl>
                         <dbl>
                                      <dbl>
#> 1
              1.4
                           0.2
                                        3.5
#> 2
              1.4
                           0.2
                                        3
#> 3
              1.3
                           0.2
                                        3.2
#> 4
              1.5
                           0.2
                                        3.1
#> # i 146 more rows
To take the difference between two selections, combine the & and ! operators:
```

```
iris %>% select(starts_with("Petal") & !ends_with("Width"))
#> # A tibble: 150 x 1
#>
     Petal.Length
#>
            <dbl>
#> 1
              1.4
#> 2
              1.4
#> 3
              1.3
#> 4
              1.5
#> # i 146 more rows
```

## References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166-1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

separate 49

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

## See Also

```
Other single table verbs: arrange(), filter(), mutate(), reframe(), rename(), slice(), summarise()
```

#### **Examples**

```
data(pasilla)
pasilla |> select(.sample, .feature, counts)
```

separate

Separate a character column into multiple columns with a regular expression or numeric locations

## **Description**

## [Superseded]

separate() has been superseded in favour of separate\_wider\_position() and separate\_wider\_delim() because the two functions make the two uses more obvious, the API is more polished, and the handling of problems is better. Superseded functions will not go away, but will only receive critical bug fixes.

Given either a regular expression or a vector of character positions, separate() turns a single character column into multiple columns.

## Usage

```
## $3 method for class 'SummarizedExperiment'
separate(
    data,
    col,
    into,
    sep = "[^[:alnum:]]+",
    remove = TRUE,
    convert = FALSE,
    extra = "warn",
    fill = "warn",
    ...
)
```

## **Arguments**

data A data frame.

col <tidy-select> Column to expand.

into Names of new variables to create as character vector. Use NA to omit the variable

in the output.

50 separate

Separator between columns. sep

> If character, sep is interpreted as a regular expression. The default value is a regular expression that matches any sequence of non-alphanumeric values.

> If numeric, sep is interpreted as character positions to split at. Positive values start at 1 at the far-left of the string; negative value start at -1 at the far-right of

the string. The length of sep should be one less than into.

If TRUE, remove input column from output data frame. remove

If TRUE, will run type.convert() with as.is = TRUE on new columns. This is convert

useful if the component columns are integer, numeric or logical.

NB: this will cause string "NA"s to be converted to NAs.

If sep is a character vector, this controls what happens when there are too many

pieces. There are three valid options:

• "warn" (the default): emit a warning and drop extra values.

• "drop": drop any extra values without a warning.

• "merge": only splits at most length(into) times

fill If sep is a character vector, this controls what happens when there are not

enough pieces. There are three valid options:

• "warn" (the default): emit a warning and fill from the right

• "right": fill with missing values on the right

• "left": fill with missing values on the left

Additional arguments passed on to methods. . . .

#### Value

tidySummarizedExperiment

## References

extra

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166-1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.Rproject.org/package=tidyr

## See Also

unite(), the complement, extract() which uses regular expression capturing groups.

#### **Examples**

```
un <- tidySummarizedExperiment::pasilla |>
    unite("group", c(condition, type))
un |> separate(col=group, into=c("condition", "type"))
```

slice 51

c	7	^	

Subset rows using their positions

## **Description**

slice() lets you index rows by their (integer) locations. It allows you to select, remove, and duplicate rows. It is accompanied by a number of helpers for common use cases:

- slice\_head() and slice\_tail() select the first or last rows.
- slice\_sample() randomly selects rows.
- slice\_min() and slice\_max() select rows with the smallest or largest values of a variable.

If . data is a grouped\_df, the operation will be performed on each group, so that (e.g.)  $slice_head(df, n = 5)$  will select the first five rows in each group.

## Usage

```
## S3 method for class 'SummarizedExperiment'
slice(.data, ..., .preserve = FALSE)
```

#### **Arguments**

. data A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

... For slice(): <data-masking> Integer row values.

Provide either positive values to keep, or negative values to drop. The values provided must be either all positive or all negative. Indices beyond the number of rows in the input are silently ignored.

For slice\_\*(), these arguments are passed on to methods.

.preserve

Relevant when the .data input is grouped. If .preserve = FALSE (the default), the grouping structure is recalculated based on the resulting data, otherwise the grouping is kept as is.

#### **Details**

Slice does not work with relational databases because they have no intrinsic notion of row order. If you want to perform the equivalent operation, use filter() and row\_number().

## Value

An object of the same type as .data. The output has the following properties:

- Each row may appear 0, 1, or many times in the output.
- · Columns are not modified.
- Groups are not modified.
- Data frame attributes are preserved.

52 summarise

#### Methods

These function are **generics**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

```
• slice(): no methods found.
```

- slice\_head(): no methods found.
- slice\_tail(): no methods found.
- slice\_min(): no methods found.
- slice\_max(): no methods found.
- slice\_sample(): no methods found.

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

```
Other single table verbs: mutate(), rename(), summarise()
```

## **Examples**

```
data(pasilla)
pasilla |> slice(1)
```

summarise

Summarise each group down to one row

## Description

summarise() creates a new data frame. It returns one row for each combination of grouping variables; if there are no grouping variables, the output will have a single row summarising all observations in the input. It will contain one column for each grouping variable and one column for each of the summary statistics that you have specified.

```
summarise() and summarize() are synonyms.
```

## Usage

```
## S3 method for class 'SummarizedExperiment'
summarise(.data, ...)
## S3 method for class 'SummarizedExperiment'
summarize(.data, ...)
```

summarise 53

## **Arguments**

.data

A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.

. . .

<data-masking> Name-value pairs of summary functions. The name will be
the name of the variable in the result.

The value can be:

- A vector of length 1, e.g. min(x), n(), or sum(is.na(y)).
- A data frame, to add multiple columns from a single expression.

[**Deprecated**] Returning values with size 0 or >1 was deprecated as of 1.1.0. Please use reframe() for this instead.

#### Value

An object usually of the same type as .data.

- The rows come from the underlying group\_keys().
- The columns are a combination of the grouping keys and the summary expressions that you provide.
- The grouping structure is controlled by the .groups= argument, the output may be another grouped\_df, a tibble or a rowwise data frame.
- Data frame attributes are not preserved, because summarise() fundamentally creates a new data frame.

## **Useful functions**

```
Center: mean(), median()
Spread: sd(), IQR(), mad()
Range: min(), max(),
Position: first(), last(), nth(),
Count: n(), n_distinct()
Logical: any(), all()
```

#### **Backend variations**

The data frame backend supports creating a variable and using it in the same summary. This means that previously created summary variables can be further transformed or combined within the summary, as in mutate(). However, it also means that summary variables with the same names as previous variables overwrite them, making those variables unavailable to later summary variables.

This behaviour may not be supported in other backends. To avoid unexpected results, consider using new names for your summary variables, especially when creating multiple summaries.

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages: no methods found.

54 tidy

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., François, R., Henry, L., Müller, K., Vaughan, D. (2023). dplyr: A Grammar of Data Manipulation. R package version 2.1.4, https://CRAN.R-project.org/package=dplyr

#### See Also

```
Other single table verbs: mutate(), rename(), slice()
```

## **Examples**

```
data(pasilla)
pasilla |> summarise(mean(counts))
```

tidy

tidy for Seurat

## **Description**

```
tidy for Seurat
```

## Usage

```
tidy(object)
## S3 method for class 'SummarizedExperiment'
tidy(object)
## S3 method for class 'RangedSummarizedExperiment'
tidy(object)
```

## **Arguments**

object

A SummarizedExperiment object

#### Value

A tidyseurat object.

unite 55

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

## **Examples**

```
data(pasilla)
pasilla %>% tidy()
```

unite

Unite multiple columns into one by pasting strings together

## **Description**

Convenience function to paste together multiple columns into one.

## Usage

```
## S3 method for class 'SummarizedExperiment'
unite(data, col, ..., sep = "_", remove = TRUE, na.rm = FALSE)
```

# Arguments

data	A data frame.	
col	The name of the new column, as a string or symbol.	
	This argument is passed by expression and supports quasiquotation (you can unquote strings and symbols). The name is captured from the expression with rlang::ensym() (note that this kind of interface where symbols do not represent actual objects is now discouraged in the tidyverse; we support it here for backward compatibility).	
	<tidy-select> Columns to unite</tidy-select>	
sep	Separator to use between values.	
remove	If TRUE, remove input columns from output data frame.	
na.rm	If TRUE, missing values will be removed prior to uniting each value.	

## Value

tidySummarizedExperiment

56 unnest

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

#### See Also

```
separate(), the complement.
```

## **Examples**

```
tidySummarizedExperiment::pasilla |>
   unite("group", c(condition, type))
```

unnest

Unnest a list-column of data frames into rows and columns

## **Description**

Unnest expands a list-column containing data frames into rows and columns.

## Usage

```
## S3 method for class 'tidySummarizedExperiment_nested'
unnest(
  data,
  cols,
  . . . ,
  keep_empty = FALSE,
  ptype = NULL,
  names_sep = NULL,
  names_repair = "check_unique",
  .drop,
  .id,
  .sep,
  .preserve
unnest_summarized_experiment(
  data,
  cols,
  keep_empty = FALSE,
  ptype = NULL,
  names_sep = NULL,
  names_repair = "check_unique",
  .drop,
  .id,
```

57 unnest

```
.sep,
.preserve
```

#### **Arguments**

data A data frame.

cols <tidy-select> List-columns to unnest.

> When selecting multiple columns, values from the same row will be recycled to their common size.

[Deprecated]: previously you could write df %>% unnest(x, y, z). Convert to df % unnest(c(x, y, z)). If you previously created a new variable in unnest() you'll now need to do it explicitly with mutate(). Convert df %>% unnest(y = fun(x, y, z)) to df % mutate(y = fun(x, y, z)) %>% unnest(y).

keep\_empty By default, you get one row of output for each element of the list that you are

> unchopping/unnesting. This means that if there's a size-0 element (like NULL or an empty data frame or vector), then that entire row will be dropped from the output. If you want to preserve all rows, use keep\_empty = TRUE to replace

size-0 elements with a single row of missing values.

Optionally, a named list of column name-prototype pairs to coerce cols to, overptype riding the default that will be guessed from combining the individual values.

Alternatively, a single empty ptype can be supplied, which will be applied to all

cols.

If NULL, the default, the outer names will come from the inner names. If a names\_sep string, the outer names will be formed by pasting together the outer and the

inner column names, separated by names\_sep.

Used to check that output data frame has valid names. Must be one of the names\_repair

following options:

- "minimal": no name repair or checks, beyond basic existence,
- "unique": make sure names are unique and not empty,
- "check\_unique": (the default), no name repair, but check they are unique,
- "universal": make the names unique and syntactic
- a function: apply custom name repair.
- tidyr legacy: use the name repair from tidyr 0.8.
- a formula: a purrr-style anonymous function (see rlang::as\_function())

See vctrs::vec\_as\_names() for more details on these terms and the strategies used to enforce them.

.drop, .preserve

[Deprecated]: all list-columns are now preserved; If there are any that you don't want in the output use select() to remove them prior to unnesting.

[Deprecated]: convert df %>% unnest(x, .id = "id") to df %>% mutate(id = names(x)) %>% unn .id

[Deprecated]: use names\_sep instead. .sep

## Value

tidySummarizedExperiment

58

#### **New syntax**

tidyr 1.0.0 introduced a new syntax for nest() and unnest() that's designed to be more similar to other functions. Converting to the new syntax should be straightforward (guided by the message you'll receive) but if you just need to run an old analysis, you can easily revert to the previous behaviour using nest\_legacy() and unnest\_legacy() as follows:

```
library(tidyr)
nest <- nest_legacy
unnest <- unnest_legacy</pre>
```

#### References

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

Hutchison, W.J., Keyes, T.J., The tidyomics Consortium. et al. The tidyomics ecosystem: enhancing omic data analyses. Nat Methods 21, 1166–1170 (2024). https://doi.org/10.1038/s41592-024-02299-2

Wickham, H., Vaughan, D. (2023). tidyr: Tidy Messy Data. R package version 2.0.0, https://CRAN.R-project.org/package=tidyr

## See Also

```
Other rectangling: hoist(), unnest_longer(), unnest_wider()
```

## **Examples**

```
tidySummarizedExperiment::pasilla |>
   nest(data=-condition) |>
   unnest(data)

tidySummarizedExperiment::pasilla |>
   nest(data=-condition) |>
   unnest_summarized_experiment(data)
```

%>%

Pipe operator

## **Description**

```
See magrittr::%>% for details.
```

## Usage

```
lhs %>% rhs
```

%>%

# Arguments

1hs A value or the magrittr placeholder.

rhs A function call using the magrittr semantics.

# Value

The result of calling rhs(lhs).

# **Examples**

```
library(magrittr)
1 %>% sum(2)
```

# Index

datasets	amagatalk, hasala() 27
* datasets	crosstalk::bscols(), 37
pasilla, 28	crosstalk::SharedData, 35
se, 44	cumal1(), 25
* internal	cumany(), 25
%>%, 58 * single table verbs	cume_dist(), 25
* single table verbs mutate, 25	$\operatorname{cummax}(), 25$
	cummean(), 25
rename, 38	cummin(), 25
slice, 51	cumsum(), 25
summarise, 52	data.frame, $3$
+, 25 .onLoad(), 4	dense_rank(), 25
** :	distinct, 8
==, 11	dplyr::group_by(), 28
>, 11	upiyi gi oup_by(), 20
>=, 11	ends_with(), <i>45</i>
?join_by, 13, 20, 23, 40	enframe(), 5
8, 11 %> % 50 50	event_data(), 36
%>%, <i>58</i> , <i>58</i>	everything(), 45
add trace() 36 37	expand(), $32$
add_trace(), 36, 37 all(), 53	extract, 9
all_of(), 45	extract(), 30, 50
	extract(), 50, 50
animation, 34	filter, 10, 49
any(), 53 $any(), 45$	filter(), 51
<pre>any_of(), 45 append_samples, 3</pre>	first(), 53
arrange, 12, 49	formula, 36
_	fortify(), <i>15</i>
arrange(), <i>17</i> as_tibble, <i>3</i>	full_join, 12
as_tibble(), 42	1 dl1_J0111, 12
as_tibble(), 42	gather(), <i>31</i>
base::as.data.frame(), 3	ggplot, 15
base::data.frame(), $3$	ggplot2::qplot(), 34
base::split(), 18	ggplotly(), 37
between(), 11	grDevices::col2rgb(), 35
bind_cols (bind_rows), 5	group_by, 16, 19
bind_rows, 5	group_by(), 7, 11, 18, 42
51114_1 5113, 5	group_by_drop_default(), 17
$case\_when(), 25$	group_cols(), 45
coalesce(), 25	group_keys(), 18, 53
contains(), 45	group_map, 18, 19
count, 7	group_nest, 18, 19
cross_join, 14, 21, 24, 41	group_split, 18, 18
cross_join(), 13, 20, 23, 40	group_split(), 18
	O:

INDEX 61

prouped_df, 17, 42, 51, 53  highlight(), 37 hoist, 58  I(), 35, 36  I(), 35, 36 if_else(), 25 inner_join, 19  IQR(), 53 is_na(), 11  join_by(), 13, 20, 22, 23, 40  lag(), 25 last(), 53 last_ol(), 45 layout(), 37 led(), 25 last_ol(), 25  mad(), 25  mad(), 25  mad(), 25  mad(), 53  matches(), 45 matrix, 3 max(), 53 matrix, 3 max(), 53 matrix, 3 max(), 53 matrix, 3 max(), 53 modian(), 53 min_rank(), 25 mutate_l(), 25, 39, 49, 52, 54 mutate(), 53  n(), 53 n_distinct(), 53 n_distinct(), 53 n_distinct(), 53 n_distinct(), 53 n_aif(), 25 nor, 11 nest_p(), 43 nest_legacy(), 27, 58 nth(), 53 nor, 36 par, 36 par, 36 par, 36 percent_rank(), 25 pivot_longer, 29 pivot_longer, 31 pivot_wider, 31 plot_wiweler_loxed plot_longer, 37 plot_ly, 37 plot_ly, 37 plot_ly, 37 plot_ly, 37 plot_longer, 37 pl	group_trim, <i>18</i> , <i>19</i>	<pre>pivot_wider_spec(), 33</pre>
highlight(), 37 hoist, 58 hoist, 58 lost, 59 lost, 59 lost, 37 lost, 37 lost, 36 lost, 58 lost, 53 lost, 54 lost, 53 lost, 53 lost, 54 los	grouped_df, 17, 42, 51, 53	plot(), <i>34</i>
hoist, 58  I(), 35, 36  if else(), 25  inner_join, 19  I(p(), 53  is.na(), 11  join_by(), 13, 20, 22, 23, 40  lag(), 25  last(), 53  last_col(), 45  layout(), 37  ledd(), 25  left_join, 22  list_of, 18  log(), 25  mad(), 53  matches(), 45  matrix, 3  max(), 53  matches(), 45  matrix, 3  max(), 53  max(), 53  max(), 53  modian(), 53  modianticol(), 53  nodistinct(), 54  nodistinct(), 54  nodistinct(), 54  nodistinct(), 57  nodistinct(), 57  nodistinct(), 57  nodistinct(), 57  nodistinct(),		plot_geo(), <i>37</i>
Diol.   Diol	highlight(), 37	plot_ly, 34
I(), 35, 36     if_else(), 25     inner_join, 19     IQR(), 33     is. na(), 11  join_by(), 13, 20, 22, 23, 40     lag(), 25     last(), 53     last_col(), 45     layout(), 37     lead(), 25     late(), 25     inner_ioin, 19  lag(), 25  last(), 53     last_col(), 45     layout(), 37     lead(), 25     left_join, 22     list_of, 18     log(), 25  mad(), 53     matrix, 3     max(), 53     median(), 53     median(), 53     median(), 53     median(), 53     median(), 53     min_rank(), 25     mutate, 12, 25, 39, 49, 52, 54     mutate(), 53     n_distinct(), 53     n_distinct(), 53     n_aif(), 25     nest_by(), 43     nest_join, 14, 21, 24, 41     nest_legacy(), 27, 58     num_range(), 45     par, 36     percent_rank(), 25     par, 36     percent_rank(), 25     par, 36     percent_rank(), 25     pivot_longer(), 31     pivot_wider, 31     tibble, 53     pivot_wider, 31     tibble, 35     pivot_wider, 31     tibble, 35     pull, 37     poply, 3     pull, 37     pull, 37     pull, 37     pull, 37     pull, 37     quasiquotation, 38, 55     recode(), 25     reframe, 12, 49     recode(), 25     reframe, 12, 49     reframe, 12, 40     sampte_ins, 12, 24, 41     slice_man(slice), 51     slice_man(slice), 51     slice_sample(slice), 51     slice_sample(), 43     slice_lad(slice), 51     slice_sample(), 43     slice_sample(), 45     summarise(), 17, 42     summarise(), 18     summarise(), 19     summarise(), 19     summaris	hoist, <i>58</i>	
1(), 3, 36 if else(), 25 inner_join, 19		
	I(), 35, 36	
inner_join, 19 IQR(), 53 is.na(), 11  join_by(), 13, 20, 22, 23, 40  lag(), 25  last(), 53  last_ol(), 45  layout(), 37  lead(), 25  mad(), 25  mad(), 53  matches(), 45  matrix, 3  matches(), 45  matrix, 3  matches(), 45  matrix, 3  matches(), 53  median(), 53  separate_wider_position(), 49  separate_wider_logition(), 49  separate_	if_else(), 25	
is.na(), 11  join_by(), 13, 20, 22, 23, 40  lag(), 25  last(), 53  last_col(), 45  layout(), 37  lead(), 25  left_join, 22  list_of, 18  log(), 25  mad(), 53  matches(), 45  matrix, 3  matrix, 3  matrix, 3  matrix, 3  matrix, 3  matrix, 53  median(), 53  min_rank(), 25  mutate, 12, 25, 39, 49, 52, 54  mutate(), 53  noistinct(), 53  noile(), 25  summarise(), 43  summarise(), 45  summarise(), 45  summarise(), 47, 42  summarise(), 17, 42  summarise(), 18  summarise(), 19  summarise(), 25  sum	inner_join, 19	puri, <i>51</i>
is.na(), II  join_by(), I3, 20, 22, 23, 40  recode(), 25  reframe, I2, 49  reframe(), 53  rename, I2, 26, 38, 49, 52, 54  right_join, 39  rlang::as_function(), 4, 57  rlang::as_function(), 4, 57  rlang::as_function(), 4, 57  rlang::ensym(), 55  row_number(), 25, 51  rowwise, 42, 53  mad(), 53  mad(), 53  matrix, 3  max(), 53  median(), 53  median(), 53  median(), 53  median(), 53  median(), 53  median(), 53  min_rank(), 25  mutate, I2, 25, 39, 49, 52, 54  mutate(), 53  n(), 53  n(), 53  n(), 53  ndistinct(), 53  ndistinct(), 53  ndistinct(), 53  na_if(), 25  near(), II  nest, 27  nest_by(), 43  nest_join, I4, 21, 24, 41  nest_legacy(), 27, 58  nth(), 53  par, 36  par, 36  parcellar and sheet sheet, 31  pivot_longer(), 31  pivot_longer(), 31  pivot_wider, 31  recode(), 25  reframe, I2, 49  reframe(), 53  reframe, I2, 26, 38, 49, 52, 54  right_join, 39  rlang::as_function(), 4, 57  rlang::as_function(), 4, 5  row.as, 42  row.as, 4  row.as, 42  row.asial	IQR(), 53	quasiquotation 38 55
Join_by(), 13, 20, 22, 23, 40   recode(), 25     lag(), 25   reframe(), 53     last_col(), 45     layout(), 37     lead(), 25     left_join, 22     list_of, 18     log(), 25     mad(), 53     matches(), 45     matrix, 3     matrix, 5     median(), 53     median(), 53     median(), 53     min_rank(), 25     mutate, 12, 25, 39, 49, 52, 54     mutate(), 53     n_distinct(), 53     n_distinct(), 55     na_if(), 25     nest_by(), 43     nest_legacy(), 27, 58     nth(), 53     nth(), 53     nest_legacy(), 27, 58     nth(), 53     nth(), 54     nth(), 55     nth(), 56     nth(), 57     nth(), 57     nth(), 58     nth(), 50     nt	is.na(), <i>11</i>	quadiquotation, 50, 55
lag(), 25 last(), 53 last_col(), 45 layout(), 37 lead(), 25 left_join, 22 list_of, 18 log(), 25 mad(), 53 matches(), 45 matrix, 3 max(), 53 median(), 53 median(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 malif(), 25 modification (), 45 modisinct(), 53 na_if(), 25 near(), 11 nest_ploy(), 43 nest_ploy(), 44 nest_ploy(), 45 nest_ploy(), 45 nest_ploy(), 45 nest_ploy(), 45 nest_ploy(), 45 nest_ploy(), 45 nest_ploy		recode() 25
lag(), 25 last(), 53 last_col(), 45 layout(), 37 lead(), 25 left_join, 22 list_of, 18 log(), 25 mad(), 53 matches(), 45 matrix, 3 median(), 53 median(), 53 median(), 53 min(), 53 min(), 53 min(), 53 min(), 53 min(), 53 modistinct(), 53 n_distinct(), 53 n_distinct(), 53 n_aif(), 25 nest_by(), 43 nest_legacy(), 27, 58 nth(), 53 nth(), 5	join_by(), 13, 20, 22, 23, 40	
lag(), 25 last(), 53 last_col(), 45 layout(), 37 lead(), 25 left_join, 22 list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 median(), 53 median(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 m_if(), 25  no(), 53 n_a_if(), 25 no(), 11 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 notile(), 25 no(), 36 percent_rank(), 25 par, 36 percent_rank(), 25 pivot_longer, 29 pivot_longer, 29 pivot_longer(), 31 pivot_wider, 31  rename, 12, 26, 38, 49, 52, 54 right_join, 39 rlang::as_function(), 4, 57 rlang::as_function()		
right_join, 39 rlast(), 35 layout(), 37 lead(), 25 left_join, 22 list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 median(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_distinct(), 53 n_distinct(), 53 n_a_if(), 25 na_if(), 25 num_range(), 45  par, 36 parcal_pin(, 45 summarise, 12, 26, 39, 49, 52, 52 summarise, 12, 26, 39, 49, 52, 54 summarise, 1	lag(), 25	
last_col(), 45 layout(), 37 lead(), 25 left_join, 22 list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 mean(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_a_if(), 25 n_a_if(), 25 near(), 11 nest_py(), 43 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 ntile(), 25 num_range(), 45 num_range(), 47 num_range(), 48 num_range(), 49 num_range(), 45 num_range(), 45 num_range(), 45 num_range(), 47 num_range(), 48 num_range(), 49 num_range(), 45 num_range(), 45 num_range(), 45 num_range(), 45 num_range(), 47 num_range(), 48 num_range(), 49 num_ra	last(), <i>53</i>	
layout(), 37 lead(), 25 left_join, 22 list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 mean(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_distinct(), 53 n_aif(), 25 near(), 11 nest_27 nest_by(), 43 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 nth(), 53 ntile(), 25 num_range(), 45 num_range(), 45 par, 36 parsilla, 28 pch, 36 percent_rank(), 25 pivot_longer, 29 pivot_longer(), 31 pivot_wider, 31  rownumes, 3, 4 rownumer(), 25 row_number(), 25 row_n	last_col(), 45	
lead(), 25 left_join, 22 list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 median(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_distinct(), 53 n_aif(), 25 near(), 11 nest, 27 nest_by(), 43 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 ntile(), 25 num_range(), 45 par, 36 percent_rank(), 25 pivot_longer, 29 pivot_longer, 29 pivot_longer(), 31 pivot_wider, 31  sample_rnac, 25 sample_n, 43 select, 12, 45 separate, 49 separate(), 10, 30, 56 separate_wider_delim(), 49 separate_wider_position(), 4 separate_wider_regex(), 9 slice_lead(slice), 51 slice_max(slice), 51 slice_sample (slice), 51 slice_sample (slice), 51 spread(), 33 starts_with(), 45, 46 style(), 37 summarise (), 17, 42 summarize (summarise), 52  table, 3 pivot_wider, 31		
left_join, 22 list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 mean(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_distinct(), 53 n_aif(), 25 near(), 11 nest_27 nest_by(), 43 nest_legacy(), 27, 58 nth(), 53 ntile(), 25 num_range(), 45 par, 36 percent_rank(), 25 par, 36 percent_rank(), 25 pivot_longer(), 31 pivot_wider, 31  rownames, 3, 4 rowwise, 42, 53  sample_frac (sample_n), 43 sample_n, 43 sample_n, 43 sample_n, 43 sample_n, 43 sample_n, 43 sample_n, 43 seperate(), 35, 37 sd(), 53 separate, 49 separate, 40 separate_wider_delim(), 49 separate_wider_regex(), 9 slice, 12, 26, 39, 49, 51, 54 slice_min (slice), 51 slice_sample(), 43 slice_tail (slice), 51 spread(), 37 summarise, 12, 26, 39, 49, 52, 52 summarise(), 17, 42 summarise, 12, 26, 39, 49, 52, 52 summarise(), 17, 42 summarise(), 18, 44		
list_of, 18 log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 median(), 53 median(), 53 median(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_distinct(), 53 n_a_if(), 25 nesr(), 11 nest_27 nest_by(), 43 nest_legacy(), 27, 58 nth(), 53 ntile(), 25 num_range(), 45  par, 36 percent_rank(), 25 par, 36 percent_rank(), 25 pivot_longer(), 31 pivot_wider, 31  sample_frac (sample_n), 43 sample_n, 43 sample_n, 43 sample_n, 43 schema(), 35, 37 sd(), 53 separate, 49 separate, wider_delim(), 49 separate_wider_delim(), 49 separate_wider_login(), 49 separate_wi		row_number(), 25, 51
log(), 25  mad(), 53 matches(), 45 matrix, 3 max(), 53 mean(), 53 mean(), 53 median(), 53 median(), 53 min(), 53 min(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53 n_distinct(), 54 n_distinct(), 53 n_distinct(), 53 n_d(), 53 n_distinct(), 53 n_d(), 53 n_distinct(), 54 n_distin		rownames, $3$ , $4$
$\begin{array}{llllllllllllllllllllllllllllllllllll$		rowwise, 42, <i>53</i>
matches(), 45 matrix, 3 max(), 53 mean(), 53 mean(), 53 median(), 53 min(), 53 min(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53  n_distinct(), 53 n_a_if(), 25 near(), 11 nest, 27 nest_by(), 43 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 ntile(), 25 num_range(), 45  par, 36 par, 36 percent_rank(), 25 pivot_longer(), 31 pivot_wider, 31  sample_n, 43 sample_n, 43 sample_n, 43 sample_n, 43 schema(), 35, 37 sd(), 53 set, 44 select, 12, 45 separate, 49 separate_wider_delim(), 49 separate_wider_position(), 4 separate_wider_regex(), 9 separate_wider_ledim(), 49 sepa	10g(), 23	
matches(), 45 matrix, 3 max(), 53 mean(), 53 median(), 53 median(), 53 min(), 53 min(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53  n_distinct(), 53 na_if(), 25 near(), 11 nest, 27 nest_by(), 43 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 ntile(), 25 num_range(), 45  par, 36 percent_rank(), 25 pivot_longer(), 31 pivot_wider, 31  sechema(), 35, 37 sd(), 53 se, 44 select, 12, 45 separate, 49 separate(), 10, 30, 56 separate_wider_delim(), 49 separate_wider_position(), 4 separate_wider_regex(), 9 slice, 12, 26, 39, 49, 51, 54 slice_head(slice), 51 slice_min(slice), 51 slice_sample(), 43 slice_tail(slice), 51 spread(), 33 starts_with(), 45, 46 style(), 37 summarise, 12, 26, 39, 49, 52, 52 summarize (summarise), 52  table, 3 tibble(), 3, 5	mad() 53	<pre>sample_frac (sample_n), 43</pre>
matrix, 3 max(), 53 mean(), 53 mean(), 53 median(), 53 min(), 53 min(), 53 min_rank(), 25 mutate, 12, 25, 39, 49, 52, 54 mutate(), 53  n_distinct(), 53  n_a_if(), 25  near(), 11 nest, 27 nest_by(), 43 nest_join, 14, 21, 24, 41 nest_legacy(), 27, 58 ntile(), 25 num_range(), 45  par, 36 percent_rank(), 25 pivot_longer(), 31 pivot_wider, 31  seched, 12, 45 separate, 49 separate(), 10, 30, 56 separate_wider_delim(), 49 separate_wider_position(), 4 separate_wider_regex(), 9 separate_wider_legacy(), 9 separate_wider_legacy(), 9 separate_wider_legacy(), 49 separate_wider_legacy(), 9 separate_wider_legacy(), 49 separate_wider_legacy(), 9 separate_wider_legacy(), 49 separate		sample_n,43
$\begin{array}{llllllllllllllllllllllllllllllllllll$		
mean(), $53$ median(), $53$ median(), $53$ min(), $53$ min_rank(), $25$ mutate, $12$ , $25$ , $39$ , $49$ , $52$ , $54$ mutate(), $53$ n_in_sin(), $53$ nest_by(), $43$ nest_join, $14$ , $21$ , $24$ , $41$ nest_legacy(), $27$ , $58$ nth(), $53$ nth(), $53$ ntile(), $25$ num_range(), $45$ par, $36$ percent_rank(), $25$ par, $36$ percent_rank(), $25$ pivot_longer(), $31$ pivot_wider, $31$ select, $12$ , $45$ separate_vider_delim(), $49$ separate_wider_delim(), $49$ slice_lega(), $4$ slice_lega(), $4$ slice_lega(), $4$ slice_lega(), $4$ slice_lega()		
median(), $53$ min(), $53$ min(), $53$ min_rank(), $25$ mutate, $12$ , $25$ , $39$ , $49$ , $52$ , $54$ mutate(), $53$ separate_wider_delim(), $49$ separate_wider_regex(), $9$ mutate(), $53$ slice, $12$ , $26$ , $39$ , $49$ , $51$ , $54$ slice_head(slice), $51$ slice_max(slice), $51$ slice_min(slice), $51$ slice_sample(slice), $51$ slice_sample(slice), $51$ slice_sample(slice), $51$ slice_sample(slice), $51$ slice_sample(), $43$ nest_join, $14$ , $21$ , $24$ , $41$ slice_tail(slice), $51$ spread(), $33$ ntile(), $25$ num_range(), $45$ supplot(), $37$ summarise(), $45$ summari		
min(1), $53$ $\min(1)$ , $53$ $\min(1$		
$\begin{array}{llllllllllllllllllllllllllllllllllll$		
mutate, $12$ , $25$ , $39$ , $49$ , $52$ , $54$ mutate(), $53$ n(), $53$ n_distinct(), $53$ na_if(), $25$ near(), $11$ nest, $27$ nest_by(), $43$ nest_legacy(), $27$ , $58$ nth(), $53$ ntile(), $25$ num_range(), $45$ par, $36$ percent_rank(), $25$ pivot_longer(), $31$ pivot_wider, $31$ separate_wider_delim(), $49$ separate_wider_position(), $49$ separate_wider_position(), $49$ separate_wider_delim(), $49$ slice_head(slice), $51$ slice_max(slice), $51$ slice_max(slice), $51$ slice_sample(slice), $51$ slice_sample(slice), $51$ slice_sample(slice), $51$ slice_sample(), $43$ slice_sample(slice), $51$ slice_sample(sli		
mutate(), $53$ separate_wider_position(), $4$ separate_wider_regex(), $9$ n(), $53$ slice, $12$ , $26$ , $39$ , $49$ , $51$ , $54$ n_distinct(), $53$ slice_head (slice), $51$ slice_max (slice), $51$ near(), $11$ slice_min (slice), $51$ nest_by(), $43$ slice_sample (slice), $51$ slice_sample(), $43$ nest_join, $14$ , $21$ , $24$ , $41$ slice_tail (slice), $51$ nest_legacy(), $27$ , $58$ nth(), $53$ ntile(), $25$ num_range(), $45$ subplot(), $37$ summarise, $12$ , $26$ , $39$ , $49$ , $52$ , $52$ par, $36$ parinage(), $45$ subplot(), $37$ summarise, $12$ , $26$ , $39$ , $49$ , $52$ , $52$ porh, $36$ percent_rank(), $25$ table, $3$ pivot_longer, $29$ tbl_df, $3$ tibble, $53$ pivot_longer(), $31$ pivot_wider, $31$ tibble(), $3$ , $5$	$min_rank(), 25$	
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	mutate, 12, 25, 39, 49, 52, 54	
$\begin{array}{llllllllllllllllllllllllllllllllllll$	mutate(), <i>53</i>	
$\begin{array}{llllllllllllllllllllllllllllllllllll$		
$\begin{array}{llllllllllllllllllllllllllllllllllll$	n(), 53	
$\begin{array}{llllllllllllllllllllllllllllllllllll$	n_distinct(), 53	
nest, 27 slice_sample (slice), 51 slice_sample(), $43$ nest_by(), $43$ nest_join, $14$ , $21$ , $24$ , $41$ slice_tail (slice), 51 nest_legacy(), $27$ , $58$ spread(), $33$ nth(), $53$ ntile(), $25$ num_range(), $45$ subplot(), $37$ subplot(), $37$ summarise, $12$ , $26$ , $39$ , $49$ , $52$ , $52$ par, $36$ summarise(), $17$ , $42$ summarise(), $17$ , $42$ summarize (summarise), $52$ prot_longer, $29$ tbl_df, $3$ pivot_longer(), $31$ tibble, $53$ pivot_wider, $31$ tibble(), $3$ , $5$	na_if(), 25	slice_max(slice), 51
$\begin{array}{llllllllllllllllllllllllllllllllllll$	near(), <i>11</i>	slice_min(slice), 51
$\begin{array}{llllllllllllllllllllllllllllllllllll$		<pre>slice_sample (slice), 51</pre>
$\begin{array}{llllllllllllllllllllllllllllllllllll$		slice_sample(), 43
$\begin{array}{llllllllllllllllllllllllllllllllllll$		slice_tail(slice), 51
$\begin{array}{lll} \operatorname{nth}(), 53 & \operatorname{starts\_with}(), 45, 46 \\ \operatorname{ntile}(), 25 & \operatorname{style}(), 37 \\ \operatorname{num\_range}(), 45 & \operatorname{subplot}(), 37 \\ \operatorname{par}, 36 & \operatorname{summarise}, 12, 26, 39, 49, 52, 52 \\ \operatorname{pasilla}, 28 & \operatorname{summarize}(\operatorname{summarise}), 52 \\ \operatorname{pch}, 36 & \operatorname{summarize}(\operatorname{summarise}), 52 \\ \operatorname{pch}, 36 & \operatorname{percent\_rank}(), 25 & \operatorname{table}, 3 \\ \operatorname{pivot\_longer}, 29 & \operatorname{tbl\_df}, 3 \\ \operatorname{pivot\_longer}(), 31 & \operatorname{tibble}, 53 \\ \operatorname{pivot\_wider}, 31 & \operatorname{tibble}(), 3, 5 \\ \end{array}$		
$\begin{array}{lll} \operatorname{ntile}(), 25 & \operatorname{style}(), 37 \\ \operatorname{num\_range}(), 45 & \operatorname{subplot}(), 37 \\ \operatorname{summarise}, 12, 26, 39, 49, 52, 52 \\ \operatorname{par}, 36 & \operatorname{summarise}(), 17, 42 \\ \operatorname{pasilla}, 28 & \operatorname{summarize}\left(\operatorname{summarise}\right), 52 \\ \operatorname{pch}, 36 & \operatorname{summarize}\left(\operatorname{summarise}\right), 52 \\ \operatorname{pch}, 36 & \operatorname{percent\_rank}(), 25 & \operatorname{table}, 3 \\ \operatorname{pivot\_longer}, 29 & \operatorname{tbl\_df}, 3 \\ \operatorname{pivot\_longer}(), 31 & \operatorname{tibble}, 53 \\ \operatorname{pivot\_wider}, 31 & \operatorname{tibble}(), 3, 5 \\ \end{array}$		• • • • • • • • • • • • • • • • • • • •
$\begin{array}{llllllllllllllllllllllllllllllllllll$		
$\begin{array}{cccccccccccccccccccccccccccccccccccc$		•
$\begin{array}{lll} & & & & & & \\ \text{pasilla, 28} & & & & & \\ \text{pch, 36} & & & & \\ \text{percent\_rank(), 25} & & & & \\ \text{pivot\_longer, 29} & & & & \\ \text{pivot\_longer(), 31} & & & \\ \text{pivot\_wider, 31} & & & \\ \end{array}$	fidii_i alige(), 43	
$\begin{array}{lll} \operatorname{pasilla,28} & \operatorname{summarize}\left(\operatorname{summarise}\right), 52 \\ \operatorname{pch}, 36 & & \\ \operatorname{percent\_rank}(), 25 & & \\ \operatorname{table}, 3 & \\ \operatorname{pivot\_longer}, 29 & & \\ \operatorname{tbl\_df}, 3 & \\ \operatorname{pivot\_longer}(), 31 & & \\ \operatorname{tibble}, 53 & \\ \operatorname{pivot\_wider}, 31 & & \\ \operatorname{tibble}(), 3, 5 & \\ \end{array}$	nar 36	
$\begin{array}{lll} & \text{pch, } 36 \\ & \text{percent\_rank(), } 25 \\ & \text{pivot\_longer, } 29 \\ & \text{pivot\_longer(), } 31 \\ & \text{pivot\_wider, } 31 \end{array} \qquad \begin{array}{ll} & \text{tibble, } 3 \\ & \text{tibble, } 53 \\ & \text{tibble(), } 3, 5 \end{array}$		
$\begin{array}{lll} \operatorname{percent\_rank}(), 25 & \operatorname{table}, 3 \\ \operatorname{pivot\_longer}, 29 & \operatorname{tbl\_df}, 3 \\ \operatorname{pivot\_longer}(), 31 & \operatorname{tibble}, 53 \\ \operatorname{pivot\_wider}, 31 & \operatorname{tibble}(), 3, 5 \end{array}$		summarize (summarise), 32
$\begin{array}{ll} pivot\_longer, 29 & tbl\_df, 3 \\ pivot\_longer(), 31 & tibble, 53 \\ pivot\_wider, 31 & tibble(), 3, 5 \end{array}$		
$\begin{array}{ccc} pivot\_longer(), 31 & tibble, 53 \\ pivot\_wider, 31 & tibble(), 3, 5 \end{array}$		
$pivot\_wider, 31$	-	
nivot wider() 29 tidy 54		
prior_mass (), 27	<pre>pivot_wider(), 29</pre>	tidy, 54

INDEX

```
tidyr_legacy, 57
ts, 3
type.convert(), 10, 50
ungroup(), 11, 42
unique.data.frame(),8
unite, 55
unite(), 50
unnest, 56
unnest_legacy(), 27, 58
unnest_longer, 58
unnest\_summarized\_experiment \ (unnest),
        56
unnest_wider, 58
vctrs::vec_as_names(), 4, 5, 30, 33, 57
where(), 45
xor(), 11
```