# Package 'sccomp'

December 2, 2025

Type Package

**Title** Differential Composition and Variability Analysis for Single-Cell Data

**Version** 2.2.0 **Date** 2024-01-15

**Description** Comprehensive R package for differential composition and variability analysis in single-cell RNA sequencing, CyTOF, and microbiome data. Provides robust Bayesian modeling with outlier detection, random effects, and advanced statistical methods for cell type proportion analysis. Features include probabilistic outlier identification, mixed-effect modeling, differential variability testing, and comprehensive visualization tools. Perfect for cancer research, immunology, developmental biology, and single-cell genomics applications.

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clear\_stan\_model\_cache

Clear Stan Model Cache

# **Description**

This function removes the Stan model cache directory and all its contents. The cache is used by cmdstanr to store compiled Stan models. Clearing the cache can be useful when:

- You're experiencing issues with cached models
- You want to force recompilation of models
- You're switching between different versions of Stan

## Usage

clear\_stan\_model\_cache(cache\_dir = sccomp\_stan\_models\_cache\_dir)

### **Arguments**

cache\_dir

A character string representing the path of the cache directory to delete. Defaults to sccomp\_stan\_models\_cache\_dir, which is the default cache location used by sccomp.

#### **Details**

The function uses unlink() with recursive = TRUE to remove the directory and all its contents. If the cache directory doesn't exist, the function will simply print a message and continue. This is a safe operation as cmdstanr will recreate the cache directory when needed.

# Value

NULL invisibly. The function is called for its side effect of removing the cache directory.

# References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

## See Also

- cmdstan\_path for information about cmdstanr's cache management
- unlink for details about the underlying file removal function

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## **Examples**

```
# Clear the default sccomp cache directory
clear_stan_model_cache()

# Clear a specific cache directory
clear_stan_model_cache("path/to/custom/cache")

# Example of when you might want to clear the cache
## Not run:

# If you're experiencing issues with model compilation
clear_stan_model_cache()
# Then try running your model again

## End(Not run)
```

counts\_obj

counts\_obj

# **Description**

A tidy example dataset containing cell counts per cell group (cluster), sample, and phenotype for differential analysis. This dataset represents the counts of cells in various phenotypes and cell groups across multiple samples.

# Usage

```
data(counts_obj)
```

#### **Format**

A tidy data frame with the following columns:

- sample: Factor, representing the sample identifier.
- type: Factor, indicating the sample type (e.g., benign, cancerous).
- $\bullet \ \ phenotype: Factor, representing the cell phenotype (e.g., B\_cell, HSC, etc.).$
- count: Integer, representing the number of cells for each cell group within each sample.
- cell\_group: Factor, representing the cell group (e.g., BM, B1, Dm, etc.).

## Value

A tibble representing cell counts per cluster, with columns for sample, type, phenotype, cell group, and counts.

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handle\_missing\_values Handle missing values in a data frame

#### **Description**

Handle missing values in a data frame

# Usage

```
handle_missing_values(data)
```

# **Arguments**

data

A data frame to process

#### Value

A data frame with missing values handled:

- For categorical variables (factor/character): NA values are converted to a factor level "NA"
- For numeric variables: NA values are replaced with the mean of the column

#### **Description**

A small example dataset containing cell counts across samples, conditions, and cell groups. This dataset is used to demonstrate the use of sccomp functions in scenarios where there is no significant difference in cell composition between conditions.

#### Usage

```
data(no_significance_df)
```

## **Format**

A tibble with the following columns:

- sample: Character. Identifier for each sample.
- condition: Character. Experimental condition or group (e.g., "X" or "Y").
- cell\_group: Character. Cell group or cell type (e.g., "A", "B").
- count: Numeric. Count of cells in the given sample, condition, and cell group.

## Value

A tibble with 34 rows and 4 columns: sample, condition, cell\_group, and count.

```
data(no_significance_df)
head(no_significance_df)
```

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plot.sccomp\_tbl

plot

#### **Description**

This function plots a summary of the results of the model.

# Usage

```
## $3 method for class 'sccomp_tbl'
plot(
    x,
    significance_threshold = 0.05,
    test_composition_above_logit_fold_change = attr(.data,
        "test_composition_above_logit_fold_change"),
    significance_statistic = c("FDR", "pH0"),
    show_fdr_message = TRUE,
    ...
)
```

#### **Arguments**

Х

A tibble including a cell\_group name column | sample name column | read counts column | factor columns | Pvalue column | a significance column

 $significance\_threshold$ 

Numeric value specifying the significance threshold for highlighting differences. Default is 0.025.

test\_composition\_above\_logit\_fold\_change

A positive integer. It is the effect threshold used for the hypothesis test. A value of 0.2 correspond to a change in cell proportion of 10% for a cell type with baseline proportion of 50%. That is, a cell type goes from 45% to 50%. When the baseline proportion is closer to 0 or 1 this effect thrshold has consistent value in the logit uncontrained scale.

significance\_statistic

Character vector indicating which statistic to highlight. Default is "FDR".

show\_fdr\_message

Logical. Whether to show the Bayesian FDR interpretation message on the plot. Default is TRUE.

... For internal use

#### Value

A ggplot

## References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

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#### **Examples**

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimate = sccomp_estimate(
        counts_obj,
        ~ type, ~1, "sample", "cell_group", "count",
        cores = 1
    )

    # estimate |> plot()
}
```

plot\_1D\_intervals

Plot 1D Intervals for Cell-group Effects

# **Description**

This function creates a series of 1D interval plots for cell-group effects, highlighting significant differences based on a given significance threshold.

#### Usage

```
plot_1D_intervals(
    .data,
    significance_threshold = 0.05,
    test_composition_above_logit_fold_change = attr(.data,
        "test_composition_above_logit_fold_change"),
    show_fdr_message = TRUE,
    significance_statistic = c("FDR", "pH0")
)
```

# **Arguments**

. data Data frame containing the main data.

 $significance\_threshold$ 

Numeric value specifying the significance threshold for highlighting differences.

 ${\tt test\_composition\_above\_logit\_fold\_change}$ 

A positive integer. It is the effect threshold used for the hypothesis test. A value of 0.2 correspond to a change in cell proportion of 10% for a cell type with baseline proportion of 50%. That is, a cell type goes from 45% to 50%. When the baseline proportion is closer to 0 or 1 this effect thrshold has consistent value in the logit uncontrained scale.

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```
show_fdr_message
```

Logical. Whether to show the Bayesian FDR interpretation message on the plot. Default is TRUE.

significance\_statistic

Character vector indicating which statistic to highlight. Default is "FDR".

#### Value

A combined plot of 1D interval plots.

## **Examples**

```
print("cmdstanr is needed to run this example.")
  if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")
    estimate <- sccomp_estimate(</pre>
      counts_obj,
      ~ type,
      ~1,
      "sample",
      "cell_group",
      "count",
      cores = 1
    ) |>
    sccomp_test()
  # Example usage:
  my_plot = plot_1D_intervals(estimate)
  }
```

plot\_2D\_intervals

Plot 2D Intervals for Mean-Variance Association

# **Description**

This function creates a 2D interval plot for mean-variance association, highlighting significant differences based on a given significance threshold.

# Usage

```
plot_2D_intervals(
    .data,
    significance_threshold = 0.05,
    test_composition_above_logit_fold_change = attr(.data,
        "test_composition_above_logit_fold_change"),
    show_fdr_message = TRUE,
    significance_statistic = c("FDR", "pH0")
)
```

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## **Arguments**

. data Data frame containing the main data.

significance\_threshold

Numeric value specifying the significance threshold for highlighting differences. Default is 0.025.

 ${\tt test\_composition\_above\_logit\_fold\_change}$ 

A positive integer. It is the effect threshold used for the hypothesis test. A value of 0.2 correspond to a change in cell proportion of 10% for a cell type with baseline proportion of 50%. That is, a cell type goes from 45% to 50%. When the baseline proportion is closer to 0 or 1 this effect thrshold has consistent value in the logit uncontrained scale.

show\_fdr\_message

Logical. Whether to show the Bayesian FDR interpretation message on the plot. Default is TRUE.

significance\_statistic

Character vector indicating which statistic to highlight. Default is "FDR".

#### Value

A ggplot object representing the 2D interval plot.

```
print("cmdstanr is needed to run this example.")
  if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")
    estimate <- sccomp_estimate(</pre>
      counts_obj,
      ~ type,
      ~type,
      "sample",
      "cell_group",
      "count",
      cores = 1
    ) |>
    sccomp_test()
  # Example usage:
  my_plot = plot_2D_intervals(estimate)
  }
```

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## **Description**

This function creates a scatterplot of cell-group proportions, optionally highlighting significant differences based on a given significance threshold.

# Usage

```
plot_scatterplot(
    .data,
    data_proportion,
    factor_of_interest,
    .cell_group,
    .sample,
    significance_threshold = 0.05,
    my_theme
)
```

### **Arguments**

A ggplot2 theme object to be applied to the plot.

Print method for sccomp objects

#### Value

my\_theme

A ggplot object representing the scatterplot.

# **Examples**

```
# Example usage:
# plot_scatterplot(.data, data_proportion, "condition", "cell_group", "sample", 0.025, theme_minimal())
```

Description

print.sccomp\_tbl

Print method for sccomp objects.

The print method for sccomp objects provides a summary of the model specifications, data dimensions, and convergence diagnostics.

The output is formatted to be easy to read and understand.

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## Usage

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

# **Arguments**

x A sccomp object

... Additional arguments passed to print

#### Value

The printed object

# **Examples**

```
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    # Create a sccomp object
    data("counts_obj")
    estimate <- sccomp_estimate(
        counts_obj,
        ~ type,
        ~1,
        "sample",
        "cell_group",
        "count",
        cores = 1
    )
    print(estimate)
}</pre>
```

sccomp\_boxplot

sccomp\_boxplot

# **Description**

Creates a boxplot visualization of the model results from sccomp. This function plots the estimated cell proportions across samples, highlighting significant changes in cell composition according to a specified factor.

# Usage

```
sccomp_boxplot(
   .data,
   factor,
   significance_threshold = 0.05,
   test_composition_above_logit_fold_change = attr(.data,
```

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```
"test_composition_above_logit_fold_change"),
remove_unwanted_effects = FALSE
)
```

#### **Arguments**

.data

A tibble containing the results from sccomp\_estimate and sccomp\_test, including the columns: cell\_group name, sample name, read counts, factor(s), p-values, and significance indicators.

factor

A character string specifying the factor of interest included in the model for stratifying the boxplot.

significance\_threshold

A numeric value indicating the False Discovery Rate (FDR) threshold for labeling significant cell-groups. Defaults to 0.05.

test\_composition\_above\_logit\_fold\_change

A positive numeric value representing the effect size threshold used in the hypothesis test. A value of 0.2 corresponds to a change in cell proportion of approximately 10% for a cell type with a baseline proportion of 50% (e.g., from 45% to 55%). This threshold is consistent on the logit-unconstrained scale, even when the baseline proportion is close to 0 or 1.

remove\_unwanted\_effects

A logical value indicating whether to remove unwanted variation from the data before plotting. Defaults to FALSE.

#### Value

A ggplot object representing the boxplot of cell proportions across samples, stratified by the specified factor.

# References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimate <- sccomp_estimate(
        counts_obj,
        formula_composition = ~ type,
        formula_variability = ~ 1,
        sample = "sample",
        cell_group = "cell_group",
        abundance = "count",
        cores = 1</pre>
```

```
) |>
sccomp_test()

# Plot the boxplot of estimated cell proportions
sccomp_boxplot(
    .data = estimate,
    factor = "type",
    significance_threshold = 0.05
)
}
```

sccomp\_calculate\_residuals

Calculate Residuals Between Observed and Predicted Proportions

# **Description**

sccomp\_calculate\_residuals computes the residuals between observed cell group proportions and the predicted proportions from a fitted sccomp model. This function is useful for assessing model fit and identifying cell groups or samples where the model may not adequately capture the observed data. The residuals are calculated as the difference between the observed proportions and the predicted mean proportions from the model.

## Usage

```
sccomp_calculate_residuals(.data)
```

#### **Arguments**

.data

A tibble of class sccomp\_tbl, which is the result of sccomp\_estimate(). This tibble contains the fitted model and associated data necessary for calculating residuals.

# **Details**

The function performs the following steps:

- 1. Extracts the predicted mean proportions for each cell group and sample using sccomp\_predict().
- 2. Calculates the observed proportions from the original count data.
- 3. Computes residuals by subtracting the predicted proportions from the observed proportions.
- 4. Returns a tibble containing the sample, cell group, residuals, and exposure (total counts per sample).

#### Value

A tibble (tbl) with the following columns:

- sample A character column representing the sample identifiers.
- cell\_group A character column representing the cell group identifiers.
- **residuals** A numeric column representing the residuals, calculated as the difference between observed and predicted proportions.
- **exposure** A numeric column representing the total counts (sum of counts across cell groups) for each sample.

#### References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

# **Examples**

```
if (instantiate::stan_cmdstan_exists() && .Platform$OS.type == "unix") {
# Load example data
data("counts_obj")
# Fit the sccomp model
estimates <- sccomp_estimate(</pre>
  counts_obj,
  formula_composition = ~ type,
  formula_variability = ~1,
  sample = "sample",
  cell_group = "cell_group",
  abundance = "count",
  approximate_posterior_inference = "all",
  cores = 1
# Calculate residuals
residuals <- sccomp_calculate_residuals(estimates)</pre>
# View the residuals
print(residuals)
}
```

sccomp\_estimate

Main Function for SCCOMP Estimate

## **Description**

The sccomp\_estimate function performs linear modeling on a table of cell counts or proportions, which includes a cell-group identifier, sample identifier, abundance (counts or proportions), and factors (continuous or discrete). The user can define a linear model using an R formula, where the first factor is the factor of interest. Alternatively, sccomp accepts single-cell data containers (e.g., Seurat, SingleCellExperiment, cell metadata, or group-size) and derives the count data from cell metadata.

## Usage

```
sccomp_estimate(
   .data,
   formula_composition = ~1,
   formula_variability = ~1,
   sample,
   cell_group,
   abundance = NULL,
```

```
cores = detectCores(),
 bimodal_mean_variability_association = FALSE,
  percent_false_positive = 5,
  inference_method = "pathfinder",
 prior_mean = list(intercept = c(0, 1), coefficients = c(0, 1)),
 prior_overdispersion_mean_association = list(intercept = c(5, 2), slope = c(0, 0.6),
    standard_deviation = c(10, 20)),
  .sample_cell_group_pairs_to_exclude = NULL,
  output_directory = "sccomp_draws_files",
  verbose = TRUE,
  enable_loo = FALSE,
  noise_model = "multi_beta_binomial",
  exclude_priors = FALSE,
 use_data = TRUE,
 mcmc_seed = sample_seed(),
 max_sampling_iterations = 20000,
  pass_fit = TRUE,
  sig_figs = 9,
  cache_stan_model = sccomp_stan_models_cache_dir,
  .count = NULL,
  approximate_posterior_inference = NULL,
  variational_inference = NULL,
  .sample = NULL,
  .cell_group = NULL,
  .abundance = NULL
)
```

# **Arguments**

.data

A tibble including cell\_group name column, sample name column, abundance column (counts or proportions), and factor columns.

formula\_composition

A formula describing the model for differential abundance.

formula\_variability

A formula describing the model for differential variability.

sample A column name as a character string for the sample identifier. Replaces the

deprecated .sample.

cell\_group A column name as a character string for the cell-group identifier. Replaces the

deprecated .cell\_group.

abundance A column name as a character string for the cell-group abundance, which can be

counts (> 0) or proportions (between 0 and 1, summing to 1 across cell\_group).

Replaces the deprecated . abundance and . count.

cores Number of cores to use for parallel calculations.

bimodal\_mean\_variability\_association

Logical, whether to model mean-variability as bimodal.

percent\_false\_positive

A real number between 0 and 100 for outlier identification.

inference\_method

Character string specifying the inference method to use ('pathfinder', 'hmc', or 'variational'). Replaces the deprecated approximate\_posterior\_inference and variational\_inference.

prior\_mean A list specifying prior knowledge about the mean distribution, including inter-

cept and coefficients.

prior\_overdispersion\_mean\_association

A list specifying prior knowledge about mean/variability association.

.sample\_cell\_group\_pairs\_to\_exclude

A column name indicating sample/cell-group pairs to exclude.

output\_directory

A character string specifying the output directory for Stan draws.

verbose Logical, whether to print progression details.

enable\_loo Logical, whether to enable model comparison using the LOO package.

noise\_model A character string specifying the noise model (e.g., 'multi\_beta\_binomial').

exclude\_priors Logical, whether to run a prior-free model.

use\_data Logical, whether to run the model data-free.

mcmc\_seed An integer seed for MCMC reproducibility.

max\_sampling\_iterations

Integer to limit the maximum number of iterations for large datasets.

pass\_fit Logical, whether to include the Stan fit as an attribute in the output.

sig\_figs Number of significant figures to use for Stan model output. Default is 9.

cache\_stan\_model

A character string specifying the cache directory for compiled Stan models. The sccomp version will be automatically appended to ensure version isolation. Default is sccomp\_stan\_models\_cache\_dir which points to ~/.sccomp\_models.

... Additional arguments passed to the cmdstanr::sample function.

. count **DEPRECATED**. Use abundance instead.

approximate\_posterior\_inference

**DEPRECATED**. Use inference\_method instead.

variational\_inference

**DEPRECATED**. Use inference\_method instead.

. sample **DEPRECATED**. Use sample instead.

. abundance **DEPRECATED**. Use abundance instead.

#### Value

A tibble (tbl), with the following columns:

- cell\_group The cell groups being tested.
- parameter The parameter being estimated from the design matrix described by the input formula\_composition and formula\_variability.
- factor The covariate factor in the formula, if applicable (e.g., not present for Intercept or contrasts).
- c\_lower Lower (2.5%) quantile of the posterior distribution for a composition (c) parameter.
- c\_effect Mean of the posterior distribution for a composition (c) parameter.
- c\_upper Upper (97.5%) quantile of the posterior distribution for a composition (c) parameter.
- c\_pHO Probability of the c\_effect being smaller or bigger than the test\_composition\_above\_logit\_fold\_change argument.

c\_FDR - False discovery rate of the c\_effect being smaller or bigger than the test\_composition\_above\_logit\_fo
argument. False discovery rate for Bayesian models is calculated differently from frequentists
models, as detailed in Mangiola et al, PNAS 2023.

- c\_n\_eff Effective sample size, the number of independent draws in the sample. The higher, the better.
- c R k hat R statistic, a measure of chain equilibrium, should be within 0.05 of 1.0.
- v\_lower Lower (2.5%) quantile of the posterior distribution for a variability (v) parameter.
- v\_effect Mean of the posterior distribution for a variability (v) parameter.
- v\_upper Upper (97.5%) quantile of the posterior distribution for a variability (v) parameter.
- v\_pHO Probability of the v\_effect being smaller or bigger than the test\_composition\_above\_logit\_fold\_chan argument.
- v\_FDR False discovery rate of the v\_effect being smaller or bigger than the test\_composition\_above\_logit\_fo argument. False discovery rate for Bayesian models is calculated differently from frequentists models, as detailed in Mangiola et al, PNAS 2023.
- v\_n\_eff Effective sample size for a variability (v) parameter.
- v\_R\_k\_hat R statistic for a variability (v) parameter, a measure of chain equilibrium.

The function also attaches several attributes to the result:

- count\_data The original count data used in the analysis, stored as an attribute for efficient access.
- model\_input The model input data used for fitting.
- formula\_composition The formula used for composition modeling.
- formula\_variability The formula used for variability modeling.
- fit The Stan fit object (if pass\_fit = TRUE).

# References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimate <- sccomp_estimate(
        counts_obj,
        ~ type,
        ~1,
        "sample",
        "cell_group",
        "count",
        cores = 1</pre>
```

18 sccomp\_predict

sccomp\_predict

sccomp\_predict

## **Description**

This function replicates counts from a real-world dataset.

# Usage

```
sccomp_predict(
  fit,
  formula_composition = NULL,
  new_data = NULL,
  number_of_draws = 500,
  mcmc_seed = sample_seed(),
  summary_instead_of_draws = TRUE,
  robust = FALSE
)
```

# **Arguments**

fit The result of sccomp\_estimate.

formula\_composition

A formula. The formula describing the model for differential abundance, for example ~treatment. This formula can be a sub-formula of your estimated model; in this case all other factor will be factored out.

new\_data

A sample-wise data frame including the column that represent the factors in your formula. If you want to predict proportions for 10 samples, there should be 10 rows. T

number\_of\_draws

An integer. How may copies of the data you want to draw from the model joint posterior distribution.

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mcmc\_seed

An integer. Used for Markov-chain Monte Carlo reproducibility. By default a random number is sampled from 1 to 999999. This itself can be controlled by set.seed()

summary\_instead\_of\_draws

Return the summary values (i.e. mean and quantiles) of the predicted proportions, or return single draws. Single draws can be helful to better analyse the uncertainty of the prediction.

robust

A logical. If TRUE, use robust statistics (median and median absolute deviation) instead of classical statistics (mean and standard deviation) for the summary calculations.

#### Value

A tibble (tbl) with the following columns:

- cell\_group A character column representing the cell group being tested.
- sample A factor column representing the sample name for which the predictions are made.
- **proportion\_mean** A numeric column representing the predicted mean (or median when robust=TRUE) proportions from the model.
- **proportion\_lower** A numeric column representing the lower bound (2.5%) of the 95% credible interval for the predicted proportions.
- **proportion\_upper** A numeric column representing the upper bound (97.5%) of the 95% credible interval for the predicted proportions.

#### References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

```
sccomp_proportional_fold_change
```

Calculate Proportional Fold Change from sccomp Estimated Effects

# **Description**

This function calculates the proportional fold change between two conditions using the estimated effects from a sccomp model. The fold changes are derived from the model's posterior predictions rather than raw counts, providing a more robust estimate that accounts for the model's uncertainty and covariate effects.

Note! This statistic is descriptive and should not be used to define significance - use sccomp\_test() for that. While fold changes in proportions are easier to interpret than changes in logit space, they are not linear (the same proportional change has different meaning for rare vs abundant cell types). In contrast, the logit scale used internally by sccomp provides linear effects that are more appropriate for statistical inference.

## Usage

```
sccomp_proportional_fold_change(.data, formula_composition, from, to)
```

# **Arguments**

.data

A sccomp estimate object (of class 'sccomp\_tbl') obtained from running sccomp\_estimate(). This object contains the fitted model and estimated effects.

formula\_composition

The formula specifying which model effects to use for calculating fold changes. This should match or be a subset of the formula used in the original sccomp\_estimate()

call

from Character string specifying the reference/control condition (e.g., "benign").

to Character string specifying the comparison condition (e.g., "cancer").

## Value

A tibble with the following columns:

- cell\_group The cell group identifier
- proportion\_fold\_change The estimated fold change in proportions between conditions. Positive values indicate increases, negative values indicate decreases.
- average\_uncertainty The average uncertainty in the fold change estimate, derived from the credible intervals
- statement A text description of the fold change, including the direction and the estimated proportions

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))
```

```
if (instantiate::stan_cmdstan_exists()) {
  # Load example data
  data("counts_obj")
  # First estimate the composition effects
  estimate <- sccomp_estimate(</pre>
      counts_obj,
      ~ type,
      ~1,
      "sample",
      "cell_group",
      "count",
      cores = 1
  )
  # Calculate proportional fold changes from the estimated effects
  estimate |>
  sccomp_proportional_fold_change(
    formula_composition = ~ type,
    from = "benign",
    to = "cancer"
 )
}
```

sccomp\_remove\_outliers

sccomp\_remove\_outliers main

#### **Description**

The sccomp\_remove\_outliers function takes as input a table of cell counts with columns for cell-group identifier, sample identifier, integer count, and factors (continuous or discrete). The user can define a linear model using an input R formula, where the first factor is the factor of interest. Alternatively, sccomp accepts single-cell data containers (e.g., Seurat, SingleCellExperiment, cell metadata, or group-size) and derives the count data from cell metadata.

# Usage

```
sccomp_remove_outliers(
    .estimate,
    percent_false_positive = 5,
    cores = detectCores(),
    inference_method = attr(.estimate, "inference_method"),
    output_directory = "sccomp_draws_files",
    verbose = TRUE,
    mcmc_seed = sample_seed(),
    max_sampling_iterations = 20000,
    enable_loo = FALSE,
    sig_figs = 9,
    cache_stan_model = sccomp_stan_models_cache_dir,
    approximate_posterior_inference = NULL,
```

```
variational_inference = NULL,
    ...
)
```

#### **Arguments**

.estimate

A tibble including a cell\_group name column, sample name column, read counts column (optional depending on the input class), and factor columns.

percent\_false\_positive

A real number between 0 and 100 (not inclusive), used to identify outliers with a specific false positive rate.

cores

Integer, the number of cores to be used for parallel calculations.

inference\_method

Character string specifying the inference method to use ('pathfinder', 'hmc', or 'variational').

output\_directory

A character string specifying the output directory for Stan draws.

verbose

Logical, whether to print progression details.

mcmc\_seed

Integer, used for Markov-chain Monte Carlo reproducibility. By default, a random number is sampled from 1 to 999999.

max\_sampling\_iterations

Integer, limits the maximum number of iterations in case a large dataset is used, to limit computation time.

enable\_loo

Logical, whether to enable model comparison using the R package LOO. This is useful for comparing fits between models, similar to ANOVA.

sig\_figs

Number of significant figures to use for Stan model output. Default is 9.

cache\_stan\_model

A character string specifying the cache directory for compiled Stan models. The sccomp version will be automatically appended to ensure version isolation. Default is sccomp\_stan\_models\_cache\_dir which points to ~/.sccomp\_models.

approximate\_posterior\_inference

DEPRECATED, use the variational\_inference argument.

variational\_inference

DEPRECATED Logical, whether to use variational Bayes for posterior inference. It is faster and convenient. Setting this argument to FALSE runs full Bayesian (Hamiltonian Monte Carlo) inference, which is slower but the gold standard.

. Additional arguments passed to the cmdstanr::sample function.

#### Value

A tibble (tbl), with the following columns:

- cell\_group The cell groups being tested.
- parameter The parameter being estimated from the design matrix described by the input formula\_composition and formula\_variability.
- factor The covariate factor in the formula, if applicable (e.g., not present for Intercept or contrasts).
- c\_lower Lower (2.5%) quantile of the posterior distribution for a composition (c) parameter.

- c\_effect Mean of the posterior distribution for a composition (c) parameter.
- c\_upper Upper (97.5%) quantile of the posterior distribution for a composition (c) parameter.
- c\_pH0 Probability of the c\_effect being smaller or bigger than the test\_composition\_above\_logit\_fold\_changer argument.
- c\_FDR False discovery rate of the c\_effect being smaller or bigger than the test\_composition\_above\_logit\_fo
  argument. False discovery rate for Bayesian models is calculated differently from frequentists
  models, as detailed in Mangiola et al, PNAS 2023.
- c\_n\_eff Effective sample size, the number of independent draws in the sample. The higher, the better.
- c\_R\_k\_hat R statistic, a measure of chain equilibrium, should be within 0.05 of 1.0.
- v\_lower Lower (2.5%) quantile of the posterior distribution for a variability (v) parameter.
- v\_effect Mean of the posterior distribution for a variability (v) parameter.
- v\_upper Upper (97.5%) quantile of the posterior distribution for a variability (v) parameter.
- v\_pHO Probability of the v\_effect being smaller or bigger than the test\_composition\_above\_logit\_fold\_chan argument.
- v\_FDR False discovery rate of the v\_effect being smaller or bigger than the test\_composition\_above\_logit\_fo argument. False discovery rate for Bayesian models is calculated differently from frequentists models, as detailed in Mangiola et al, PNAS 2023.
- v\_n\_eff Effective sample size for a variability (v) parameter.
- v\_R\_k\_hat R statistic for a variability (v) parameter, a measure of chain equilibrium.

The function also attaches several attributes to the result:

- count\_data The original count data used in the analysis, stored as an attribute for efficient access.
- model input The model input data used for fitting.
- formula\_composition The formula used for composition modeling.
- formula\_variability The formula used for variability modeling.
- fit The Stan fit object (if pass\_fit = TRUE).

#### References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimate = sccomp_estimate(
        counts_obj,
        ~ type,
```

```
~1,
    "sample",
    "cell_group",
    "count",
    cores = 1
) |>
    sccomp_remove_outliers(cores = 1)
}
```

sccomp\_remove\_unwanted\_effects

Remove unwanted effects from a sccomp\_tbl object

#### **Description**

This method removes unwanted effects from a dataset using the model estimates. For example, if you fit your data with the formula ~ factor\_1 + factor\_2 and use the formula ~ factor\_1 to remove unwanted variation, the factor\_2 effect will be factored out.

## Usage

```
sccomp_remove_unwanted_effects(
   .data,
   formula_composition_keep = NULL,
   formula_composition = NULL,
   cores = detectCores()
)
```

## **Arguments**

 $. \, data \qquad \qquad A \,\, sccomp\_tbl \,\, object. \,\, The \,\, result \,\, of \,\, sccomp\_estimate. \\ formula\_composition\_keep$ 

A formula. The formula describing the model for differential abundance, for example ~type. In this case, only the effect of the type factor will be preserved, while all other factors will be factored out.

 $formula\_composition$ 

DEPRECATED. Use formula\_composition\_keep instead.

cores Integer, the number of cores to be used for parallel calculations.

#### Value

A tibble (tbl) with the following columns:

- sample A character column representing the sample name for which data was adjusted.
- **cell\_group** A character column representing the cell group being tested.
- adjusted\_proportion A numeric column representing the adjusted proportion after removing unwanted variation.
- adjusted\_counts A numeric column representing the adjusted counts after removing unwanted variation.
- logit\_residuals A numeric column representing the logit residuals calculated after adjustment.

#### References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

# **Examples**

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimates = sccomp_estimate(
        counts_obj,
        ~ type, ~1, "sample", "cell_group", "count",
        cores = 1
    ) |>
        sccomp_remove_unwanted_effects()
```

sccomp\_remove\_unwanted\_variation

DEPRECATED: Remove Unwanted Variation from sccomp Estimates

# Description

This function is DEPRECATED. Please use sccomp\_remove\_unwanted\_effects instead. This function uses the model to remove unwanted variation from a dataset using the estimates of the model. For example, if you fit your data with the formula ~ factor\_1 + factor\_2 and use the formula ~ factor\_1 to remove unwanted variation, the factor\_2 effect will be factored out.

# Usage

```
sccomp_remove_unwanted_variation(
   .data,
   formula_composition_keep = NULL,
   formula_composition = NULL,
   formula_variability = NULL,
   cores = detectCores()
)
```

# **Arguments**

.data A tibble. The result of sccomp\_estimate.

```
formula_composition_keep
```

A formula. The formula describing the model for differential abundance, for example ~type. In this case, only the effect of the type factor will be preserved, while all other factors will be factored out.

formula\_composition

DEPRECATED. Use formula\_composition\_keep instead.

formula\_variability

DEPRECATED. Use formula\_variability\_keep instead.

cores

Integer, the number of cores to be used for parallel calculations.

#### Value

A tibble (tb1) with the following columns:

- sample A character column representing the sample name for which data was adjusted.
- cell\_group A character column representing the cell group being tested.
- adjusted\_proportion A numeric column representing the adjusted proportion after removing unwanted variation.
- adjusted\_counts A numeric column representing the adjusted counts after removing unwanted variation.
- logit\_residuals A numeric column representing the logit residuals calculated after adjustment

## References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimates = sccomp_estimate(
        counts_obj,
        ~ type, ~1, "sample", "cell_group", "count",
        cores = 1
    ) |>
        sccomp_remove_unwanted_variation()
```

sccomp\_replicate 27

sccomp\_replicate

sccomp\_replicate

## **Description**

This function replicates counts from a real-world dataset.

#### Usage

```
sccomp_replicate(
   fit,
   formula_composition = NULL,
   formula_variability = NULL,
   number_of_draws = 1,
   mcmc_seed = sample_seed(),
   cache_stan_model = sccomp_stan_models_cache_dir
)
```

# **Arguments**

fit The result of sccomp\_estimate.

formula\_composition

A formula. The formula describing the model for differential abundance, for example ~treatment. This formula can be a sub-formula of your estimated model; in this case all other factor will be factored out.

formula\_variability

A formula. The formula describing the model for differential variability, for example ~treatment. In most cases, if differentially variability is of interest, the formula should only include the factor of interest as a large amount of data is needed to define variability depending to each factors. This formula can be a sub-formula of your estimated model; in this case all other factor will be factored out.

number\_of\_draws

An integer. How may copies of the data you want to draw from the model joint posterior distribution.

mcmc\_seed

An integer. Used for Markov-chain Monte Carlo reproducibility. By default a random number is sampled from 1 to 999999. This itself can be controlled by set.seed()

cache\_stan\_model

A character string specifying the cache directory for compiled Stan models. The sccomp version will be automatically appended to ensure version isolation. Default is sccomp\_stan\_models\_cache\_dir which points to ~/.sccomp\_models.

#### Value

A tibble tbl with cell\_group-wise statistics

A tibble (tbl), with the following columns:

- cell\_group A character column representing the cell group being tested.
- sample A factor column representing the sample name from which data was generated.

- generated\_proportions A numeric column representing the proportions generated from the model.
- generated\_counts An integer column representing the counts generated from the model.
- **replicate** An integer column representing the replicate number, where each row corresponds to a different replicate of the data.

#### References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

## **Examples**

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists() && .Platform$OS.type == "unix") {
    data("counts_obj")

    sccomp_estimate(
        counts_obj,
        ~ type, ~1, "sample", "cell_group", "count",
        cores = 1
    ) |>
        sccomp_replicate()
}
```

```
sccomp_stan_models_cache_dir
```

Default cache directory for Stan models

# **Description**

A global variable that defines the default cache directory for Stan models used by the sccomp package. This directory is used to store compiled Stan models to avoid recompilation on subsequent runs.

# Usage

```
sccomp_stan_models_cache_dir
```

# **Format**

An object of class character of length 1.

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#### **Details**

The cache directory is set to ~/.sccomp\_models by default. This location is used by various sccomp functions to store and retrieve compiled Stan models, improving performance by avoiding unnecessary recompilation of models that have already been compiled.

Users can override this default by specifying a different cache directory in function calls that accept a cache\_stan\_model parameter.

#### Value

A character string containing the path to the default cache directory.

## See Also

```
sccomp_estimate, sccomp_replicate, clear_stan_model_cache
```

### **Examples**

```
# View the default cache directory
sccomp_stan_models_cache_dir

# Use a custom cache directory in a function call
# sccomp_estimate(data, cache_stan_model = "/path/to/custom/cache")
```

 $sccomp\_test$ 

sccomp\_test

# Description

This function test contrasts from a sccomp result.

# Usage

```
sccomp_test(
   .data,
   contrasts = NULL,
   percent_false_positive = 5,
   test_composition_above_logit_fold_change = 0.1,
   pass_fit = TRUE
)
```

# **Arguments**

.data

A tibble. The result of sccomp\_estimate.

contrasts

A vector of character strings. For example if your formula is  $\sim 0$  + treatment and the factor treatment has values yes and no, your contrast could be "constrasts = c(treatmentyes - treatmentno)".

percent\_false\_positive

A real between 0 and 100 non included. This used to identify outliers with a specific false positive rate.

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test\_composition\_above\_logit\_fold\_change

A positive integer. It is the effect threshold used for the hypothesis test. A value of 0.2 correspond to a change in cell proportion of 10% for a cell type with baseline proportion of 50%. That is, a cell type goes from 45% to 50%. When the baseline proportion is closer to 0 or 1 this effect thrshold has consistent value in the logit uncontrained scale.

pass\_fit

A boolean. Whether to pass the Stan fit as attribute in the output. Because the Stan fit can be very large, setting this to FALSE can be used to lower the memory imprint to save the output.

#### Value

A tibble (tbl), with the following columns:

- cell\_group The cell groups being tested.
- parameter The parameter being estimated from the design matrix described by the input formula\_composition and formula\_variability.
- factor The covariate factor in the formula, if applicable (e.g., not present for Intercept or contrasts).
- c\_lower Lower (2.5%) quantile of the posterior distribution for a composition (c) parameter.
- c\_effect Mean of the posterior distribution for a composition (c) parameter.
- c\_upper Upper (97.5%) quantile of the posterior distribution for a composition (c) parameter.
- c\_pH0 Probability of the c\_effect being smaller or bigger than the test\_composition\_above\_logit\_fold\_changer argument.
- c\_FDR False-discovery rate of the null hypothesis (no difference) for a composition (c).
- c\_n\_eff Effective sample size the number of independent draws in the sample, the higher the better.
- c\_R\_k\_hat R statistic, a measure of chain equilibrium, should be within 0.05 of 1.0.
- v\_lower Lower (2.5%) quantile of the posterior distribution for a variability (v) parameter.
- v\_effect Mean of the posterior distribution for a variability (v) parameter.
- v upper Upper (97.5%) quantile of the posterior distribution for a variability (v) parameter.
- v\_pH0 Probability of the null hypothesis (no difference) for a variability (v).
- v\_FDR False-discovery rate of the null hypothesis (no difference) for a variability (v).
- v\_n\_eff Effective sample size for a variability (v) parameter.
- v R k hat R statistic for a variability (v) parameter.
- count\_data Nested input count data.

#### References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

sccomp\_theme 31

## **Examples**

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")

    estimates = sccomp_estimate(
        counts_obj,
        ~ 0 + type, ~1, "sample", "cell_group", "count",
        cores = 1
    ) |>
        sccomp_test("typecancer - typebenign")
}
```

 $sccomp\_theme$ 

Create a multipanel theme compatible with ggplot2 4.0.0 S7 system

## **Description**

Creates a multipanel theme that is compatible with the new ggplot2 4.0.0 S7 system. This function creates a fresh theme object using the new S7 system instead of relying on the saved S3 theme object.

# Usage

```
sccomp_theme()
```

# Value

A ggplot2 theme object compatible with ggplot2 4.0.0

sce\_obj

sce\_obj

# **Description**

Example SingleCellExperiment object containing gene expression data for 106,297 cells across two assays: counts and logcounts. The object includes metadata and assay data for RNA expression, which can be used directly in differential analysis functions like sccomp\_glm.

# Usage

```
data(sce_obj)
```

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#### **Format**

A SingleCellExperiment object with the following structure:

- assays: Two assays: counts (raw RNA counts) and logcounts (log-transformed counts).
- rowData: No additional row-level metadata is present.
- colData: Metadata for each cell, including six fields: sample, type, nFeature\_RNA, ident, and others.
- dim: 1 feature and 106,297 cells.
- colnames: Cell identifiers for all 106,297 cells.

#### Value

A SingleCellExperiment object containing single-cell RNA expression data.

seurat\_obj seurat\_obj

# **Description**

Example Seurat object containing gene expression data for 106,297 cells across a single assay. The object includes RNA counts and data layers, but no variable features are defined. This dataset can be directly used with functions like sccomp\_glm for differential abundance analysis.

# Usage

```
data(seurat_obj)
```

### **Format**

A Seurat object with the following structure:

- assays: Contains gene expression data. The active assay is RNA, with 1 feature and no variable features.
- layers: Two layers: counts and data, representing raw and processed RNA expression values, respectively.
- samples: 106,297 samples (cells) within the RNA assay.

# Value

A Seurat object containing single-cell RNA expression data.

simulate\_data 33

simulate\_data

simulate\_data

#### **Description**

This function simulates data from a fitted model.

## Usage

```
simulate_data(
    .data,
    .estimate_object,
    formula_composition,
    formula_variability = NULL,
    .sample = NULL,
    .cell_group = NULL,
    .coefficients = NULL,
    variability_multiplier = 5,
    number_of_draws = 1,
    mcmc_seed = sample_seed(),
    cores = detectCores(),
    sig_figs = 9,
    cache_stan_model = sccomp_stan_models_cache_dir
)
```

## **Arguments**

.data

A tibble including a cell\_group name column | sample name column | read counts column | factor columns | Pvalue column | a significance column

.estimate\_object

The result of sccomp\_estimate execution. This is used for sampling from real-data properties.

formula\_composition

A formula. The formula describing the model for differential abundance, for example ~treatment

formula\_variability

A formula. The formula describing the model for differential variability, for example ~treatment

. sample A column name as symbol. The sample identifier

.cell\_group A column name as symbol. The cell\_group identifier

. coefficients  $\,$  The column names for coefficients, for example,  $c(b\_0,\,b\_1)$ 

variability\_multiplier

A real scalar. This can be used for artificially increasing the variability of the simulation for benchmarking purposes.

number\_of\_draws

An integer. How may copies of the data you want to draw from the model joint posterior distribution.

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mcmc\_seed An integer. Used for Markov-chain Monte Carlo reproducibility. By default a

random number is sampled from 1 to 999999. This itself can be controlled by set.seed()#' @param cores Integer, the number of cores to be used for parallel

calculations.

cores Integer, the number of cores to be used for parallel calculations.

sig\_figs Number of significant figures to use for Stan model output. Default is 9.

cache\_stan\_model

A character string specifying the cache directory for compiled Stan models. The sccomp version will be automatically appended to ensure version isolation. Default is sccomp\_stan\_models\_cache\_dir which points to ~/.sccomp\_models.

#### Value

A tibble (tbl) with the following columns:

- sample A character column representing the sample name.
- type A factor column representing the type of the sample.
- phenotype A factor column representing the phenotype in the data.
- count An integer column representing the original cell counts.
- **cell\_group** A character column representing the cell group identifier.
- **b\_0** A numeric column representing the first coefficient used for simulation.
- **b\_1** A numeric column representing the second coefficient used for simulation.
- **generated\_proportions** A numeric column representing the generated proportions from the simulation.
- generated\_counts An integer column representing the generated cell counts from the simulation.
- replicate An integer column representing the replicate number for each draw from the posterior distribution.

## References

S. Mangiola, A.J. Roth-Schulze, M. Trussart, E. Zozaya-Valdés, M. Ma, Z. Gao, A.F. Rubin, T.P. Speed, H. Shim, & A.T. Papenfuss, sccomp: Robust differential composition and variability analysis for single-cell data, Proc. Natl. Acad. Sci. U.S.A. 120 (33) e2203828120, https://doi.org/10.1073/pnas.2203828120 (2023).

```
print("cmdstanr is needed to run this example.")
# Note: Before running the example, ensure that the 'cmdstanr' package is installed:
# install.packages("cmdstanr", repos = c("https://stan-dev.r-universe.dev/", getOption("repos")))

if (instantiate::stan_cmdstan_exists()) {
    data("counts_obj")
    library(dplyr)

    estimate = sccomp_estimate(
        counts_obj,
        ~ type, ~1, "sample", "cell_group", "count",
        cores = 1
```

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```
# Set coefficients for cell_groups. In this case all coefficients are 0 for simplicity.
counts_obj = counts_obj |> mutate(b_0 = 0, b_1 = 0)

# Simulate data
simulate_data(counts_obj, estimate, ~type, ~1, sample, cell_group, c(b_0, b_1))
}
```

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