

# Package: bbsBayes2 (via r-universe)

August 26, 2024

**Type** Package

**Title** Hierarchical Bayesian Analysis of North American BBS Data

**Version** 1.1.1

**Imports** backports ( $\geq 1.4.1$ ), dplyr ( $\geq 1.1.0$ ), ggrepel, geofacet, ggplot2 ( $\geq 3.4.0$ ), HDInterval, magrittr, mgcv, posterior ( $\geq 1.2.1$ ), purrr ( $\geq 0.3.4$ ), readr ( $\geq 2.1.2$ ), rlang ( $\geq 0.4.11$ ), sbtools, sf ( $\geq 1.0.8$ ), spdep ( $\geq 1.2.7$ ), snakecase, stringr, tidyr ( $\geq 1.2.0$ ), units ( $\geq 0.8.0$ ), withr ( $\geq 2.5.0$ )

**Depends** R ( $\geq 3.5$ )

**SystemRequirements** CmdStan

(<https://mc-stan.org/users/interfaces/cmdstan>)

**URL** <https://github.com/bbsBayes/bbsBayes2>,

<https://bbsbayes.github.io/bbsBayes2/>

**BugReports** <https://github.com/bbsBayes/bbsBayes2/issues>

**NeedsCompilation** no

**Description** The North American Breeding Bird Survey (BBS) is a long-running program that seeks to monitor the status and trends of the breeding birds in North America. Since its start in 1966, the BBS has accumulated over 50 years of data for over 500 species of North American Birds. Given the temporal and spatial structure of the data, hierarchical Bayesian models are used to assess the status and trends of these 500+ species of birds. 'bbsBayes2' allows you to perform hierarchical Bayesian analysis of BBS data. You can run a full model analysis for one or more species that you choose, or you can take more control and specify how the data should be stratified, prepared for 'Stan', or modelled.

**License** MIT + file LICENSE

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<https://ropensci.r-universe.dev>  
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**Repository** <https://bbsbayes.r-universe.dev>  
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## Contents

assign_prov_state . . . . .	3
bbsBayes2-defunct . . . . .	4
bbsBayes2-deprecated . . . . .	4
bbs_data_sample . . . . .	5
bbs_models . . . . .	5
bbs_strata . . . . .	6
copy_model_file . . . . .	7
fetch_bbs_data . . . . .	8
generate_indices . . . . .	9
generate_trends . . . . .	12
get_convergence . . . . .	15
get_model_vars . . . . .	16
get_summary . . . . .	17
have_bbs_data . . . . .	18
have_cmdstan . . . . .	19
load_bbs_data . . . . .	19
load_map . . . . .	20
pacific_wren_model . . . . .	21
plot_geofacet . . . . .	22
plot_indices . . . . .	23
plot_map . . . . .	25
prepare_data . . . . .	26
prepare_model . . . . .	28
prepare_spatial . . . . .	30
remove_cache . . . . .	32
run_model . . . . .	33
save_model_run . . . . .	35
search_species . . . . .	36
species_forms . . . . .	37
stratify . . . . .	38

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assign_prov_state	<i>Categorize polygon by Province/State</i>
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### Description

Categorizes custom stratification polygons by province or state if possible. This can be useful for calculating regional indices (`generate_indices()`) or trends (`generate_trends()`) on a custom stratification, or if you want to create a geofaceted plot (`plot_geofacet()`).

### Usage

```
assign_prov_state(  
  strata_map,  
  min_overlap = 0.75,  
  plot = FALSE,  
  keep_spatial = TRUE  
)
```

### Arguments

<code>strata_map</code>	sf data frame. Strata polygons to be categorized.
<code>min_overlap</code>	Numeric. The minimum proportion of overlap between a stratum polygon and a Province or State. Overlap below this proportion will raise warnings.
<code>plot</code>	Logical. Whether to plot how polygons were assigned to Provinces or States
<code>keep_spatial</code>	Logical. Whether the output should be a spatial data frame or not.

### Value

(Spatial) data frame with strata assigned to Province/State.

### See Also

Other helper functions: [load\\_map\(\)](#), [search\\_species\(\)](#)

### Examples

```
# Demonstration of why we can't divide BCR by Provinces and States!  
map <- load_map("bcr")  
assign_prov_state(map, plot = TRUE)  
  
# Use custom stratification, using sf map object  
# e.g. with WBPHS stratum boundaries 2019  
# available: https://ecos.fws.gov/ServCat/Reference/Profile/142628  
  
## Not run:  
map <- sf::read_sf("../WBPHS_Stratum_Boundaries_2019") %>%  
  rename(strata_name = STRAT) # expects this column
```

```
s <- assign_prov_state(map, plot = TRUE)
# Some don't divide nicely, we could try a different min_overlap

s <- assign_prov_state(map, min_overlap = 0.6, plot = TRUE)

## End(Not run)
```

---

bbsBayes2-defunct      *bbsBayes2 defunct functions*

---

### **Description**

**Superseded:**

No superseded functions

**No longer relevant:**

No non-relevant functions

### **Arguments**

...                      Original function arguments

### **See Also**

[bbsBayes2-deprecated](#)

---

bbsBayes2-deprecated      *bbsBayes2 deprecated functions*

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### **Description**

No deprecated functions

### **Arguments**

...                      Original function arguments

### **See Also**

[bbsBayes2-defunct](#)

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bbs_data_sample	<i>Sample BBS data</i>
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**Description**

Contains only Pacific Wren data

**Usage**

bbs\_data\_sample

**Format**

bbs\_data\_sample:

A list containing:

- birds - counts of each bird seen per route per
- routes - data for each route run per year
- species - species list of North America

**Details**

A sample dataset containing only data for Pacific Wrens for the 2022 state-level BBS data. The full count set is obtained via the function `fetch_bbs_data()`. The data is obtained from the United States Geological Survey and is subject to change as new data is added each year. See References for citation.

**Source**

<https://www.sciencebase.gov/> via `fetch_bbs_data()`

**References**

Ziolkowski Jr., D.J., Lutmerding, M., Aponte, V.I., and Hudson, M-A.R., 2022, North American Breeding Bird Survey Dataset 1966 - 2021: U.S. Geological Survey data release, <https://doi.org/10.5066/P97WAZE5>

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bbs_models	<i>Stan models included in bbsBayes2</i>
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**Description**

These models are included in `bbsBayes2`. The model files themselves can be found in the folder identified by `system.file("models", package = "bbsBayes2")`. To create a custom Stan model, see `copy_model_file()` and the `model_file` argument of `prepare_model()`. See also the [models article](#) for more details.

**Usage**

```
bbs_models
```

**Format**

`bbs_models`:

A data frame with 9 rows and 3 columns:

- `model` - Model type
  - `first_diff` - First difference models
  - `gam` - General Additive Models (GAM)
  - `gamye` - General Additive Models (GAM) with Year Effect
  - `slope` - Slope models
- `variant` - Variant of the model to run
  - `nonhier` - Non-hierarchical models (only available for first difference models)
  - `hier` - Hierarchical models
  - `spatial` - Spatial models
- `file` - Stan model file name

**Examples**

```
bbs_models
```

---

<code>bbs_strata</code>	<i>List of included strata</i>
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---

**Description**

List of strata included in `bbsBayes2`. Each list item contains a data frame describing the strata for that stratification (name, area, country, etc.)

**Usage**

```
bbs_strata
```

**Format**

`bbs_strata`:

A list of 5 data frames

Contains `bbs_usgs`, `bbs_cws`, `bcr`, `latlong` and `prov_state`

**Examples**

```
bbs_strata[["bbs_cws"]]
bbs_strata[["latlon"]]
```

---

copy_model_file	<i>Copy model file</i>
-----------------	------------------------

---

## Description

Save a predefined Stan model file to a local text file for editing. These files can then be used in `prepare_model()` by specifying the `model_file` argument.

## Usage

```
copy_model_file(model, model_variant, dir, overwrite = FALSE)
```

## Arguments

<code>model</code>	Character. Type of model to use, must be one of "first_diff" (First Differences), "gam" (General Additive Model), "gamy" (General Additive Model with Year Effect), or "slope" (Slope model).
<code>model_variant</code>	Character. Model variant to use, must be one of "nonhier" (Non-hierarchical), "hier" (Hierarchical; default), or "spatial" (Spatially explicit).
<code>dir</code>	Character. Directory where file should be saved.
<code>overwrite</code>	Logical. Whether to overwrite an existing copy of the model file.

## Value

File path to copy of the new model file.

## See Also

Other modelling functions: [run\\_model\(\)](#), [save\\_model\\_run\(\)](#)

## Examples

```
# Save the Slope model in temp directory
copy_model_file(model = "slope", model_variant = "spatial", dir = tempdir())

# Overwrite an existing copy
copy_model_file(model = "slope", model_variant = "spatial", dir = tempdir(),
               overwrite = TRUE)

# Clean up
unlink(file.path(tempdir(), "slope_spatial_bbs_CV_COPY.stan"))
```

---

 fetch\_bbs\_data

*Fetch Breeding Bird Survey dataset*


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### Description

Fetch and download Breeding Bird Survey data from the United States Geological Survey (USGS) FTP site. This is the raw data that is uploaded to the site before any analyses are performed. Users can download different types (`state`, `stop`) and different releases (currently 2020, 2022, and 2023).

### Usage

```
fetch_bbs_data(
  level = "state",
  release = 2023,
  force = FALSE,
  quiet = FALSE,
  compression = "none"
)
```

### Arguments

<code>level</code>	Character. Which type of BBS counts to use, "state" or "stop". Default "state".
<code>release</code>	Numeric. Which yearly release to use, 2022 (including data through 2021 field season) or 2020 (including data through 2019). Default 2022.
<code>force</code>	Logical. Should pre-existing BBS data be overwritten? Default FALSE.
<code>quiet</code>	Logical. Suppress progress messages? Default FALSE.
<code>compression</code>	Character. What compression should be used to save data? Default is none which takes up the most space but is the fastest to load. Must be one of none, gz, bz2, or xz (passed to <code>readr::write_rds()</code> 's <code>compress</code> argument).

### Details

Users will be asked before saving the BBS data to a package-specific directory created on their computer. Before downloading any data, users must thoroughly read through the USGS terms and conditions for that data and enter the word "yes" to agree.

BBS state level counts provide counts beginning in 1966, aggregated in five bins, each of which contains cumulative counts from 10 of the 50 stops along a route. In contrast BBS stop level counts provides stop-level data beginning in 1997, which includes counts for each stop along routes individually. **Note that stop-level data is not currently supported by the modelling utilities in `bbsBayes2`.**

There are three releases for each type of data, 2020, 2022, and 2023. By default all functions use the most recent release unless otherwise specified. For example, the `release` argument in `stratify()` can be changed to 2020 to use the 2020 release of state-level counts.



**See Also**

Other BBS data functions: [have\\_bbs\\_data\(\)](#), [load\\_bbs\\_data\(\)](#), [remove\\_cache\(\)](#)

**Examples**

```
fetch_bbs_data(force = TRUE)
fetch_bbs_data(level = "stop", force = TRUE)
fetch_bbs_data(release = 2020, force = TRUE)
fetch_bbs_data(release = 2020, level = "stop", force = TRUE)
```

---

generate_indices	<i>Regional annual indices of abundance</i>
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**Description**

Calculate annual indices of relative abundance by year for different regions. These indices can then be used to plot population trajectories for the species, and to estimate trends.

**Usage**

```
generate_indices(
  model_output = NULL,
  quantiles = c(0.025, 0.05, 0.25, 0.75, 0.95, 0.975),
  regions = c("continent", "stratum"),
  regions_index = NULL,
  alternate_n = "n",
  start_year = NULL,
  max_backcast = NULL,
  drop_exclude = FALSE,
  hpdi = FALSE,
  quiet = FALSE
)
```

**Arguments**

model_output	List. Model output generated by <code>run_model()</code> .
quantiles	Numeric. Vector of quantiles to be sampled from the posterior distribution. Default is <code>c(0.025, 0.05, 0.25, 0.5, 0.75, 0.95, 0.975)</code> . Note that these quantiles will be used to create confidence interval bands in <code>plot_indices()</code> and by quantiles in <code>generate_trends()</code> , so make sure you specify the ones you want to use later.
regions	Character. Which region(s) to summarize and calculate indices for. Default is "continent" and "stratum". Options also include "country", "prov_state", "bcr", and "bcr_by_country". Note that some regions only apply to specific stratifications. You can also supply a custom region that exists as a column in the <code>regions_index</code> data frame (see examples for more details).

<code>regions_index</code>	Data frame. Custom regions to summarize. Data frame must include all strata in the original data in one column ( <code>strata_name</code> ), and any custom regions defined as categories in other columns. See examples for more details.
<code>alternate_n</code>	Character. Indicating the name of the alternative annual index parameter in a model, Default is "n" which for all models represents an index of the estimated annual relative abundance, scaled as the expected mean count averaged across all BBS routes and observers in a given stratum and year. For some of the models included in <code>bbsBayes2</code> , alternatives exist that provide a partial decomposition of the time-series. For the "gamye" models, the parameter "n_smooth" represents the smooth-only version of the annual index of relative abundance (i.e., the component of the annual index estimated by the spline-based smooth of the GAM). This "n_smooth" is identical to the "n" values for the same model, but excludes the annual fluctuations. For the "gamye" models, this "n_smooth" parameter is likely the most natural parameter to use in estimating trends. A similar option exists for the "slope" models, where the parameter "n_slope" represents the component of the population trajectory estimated by the log-linear regression slope parameters in the model. Users should be particularly cautious about interpreting this "n_slope" values for relatively long time-series. As a continuous regression slope, it assumes interpreting it as an estimate of population trajectory and using it to generate trend estimates assumes that there is a single continuous rate of population change across the entire time-series. Biologically, this may be reasonable for 10-20 year periods, but will be less reasonable for longer time-periods.
<code>start_year</code>	Numeric. Trim the data record before calculating annual indices.
<code>max_backcast</code>	Numeric. The number of years to back cast stratum-level estimates before the first year that species was observed on any route in that stratum. Default is NULL, which generates annual indices for the entire time series and ignores back-casting. CWS national estimates use a back cast of 5. Note that unless <code>drop_exclude = TRUE</code> , problematic years are only flagged, not omitted. See Details for more specifics.
<code>drop_exclude</code>	Logical. Whether or not strata that exceed the <code>max_backcast</code> threshold should be excluded from the calculations. Default is FALSE (regions are flagged and listed but not dropped).
<code>hpd</code>	Logical. Should credible intervals and limits be calculated using highest posterior density intervals instead of simple quantiles of the posterior distribution. Default is FALSE. these intervals are often a better descriptor of skewed posterior distributions, such as the predicted mean counts that the indices represent. Note hpd intervals are not stable for small percentages of the posterior distribution, and so <code>hpd = TRUE</code> is ignored for quantiles values between 0.33 and 0.67 (i.e., if the quantiles value defines a limit for a centered hpd interval that would include < 33% of the posterior distribution).
<code>quiet</code>	Logical. Suppress progress messages? Default FALSE.

### Details

`max_backcast` is a way to deal with the fact that the species of interest may not appear in the data until several years after the start of the time-series `max_backcast` specifies how many years can

occur before the stratum is flagged. A `max_backcast` of 5 will flag any stratum without a non-zero (or non-NA) observation within the first 5 years of the time-series. Note that records are *only* flagged unless `drop_exclude = TRUE`. If you find that the early data record is sparse and results in the exclusion of many strata, consider trimming the early years by specifying a `start_year`.

## Value

A list containing

- `indices` - data frame of calculated regional annual indices of abundances
- `samples` - array of posterior draws from the model
- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata meta data
- `raw_data` - data frame of summarized counts

`indices` contains the following columns:

- `year` - Year of particular index
- `region` - Region name
- `region_type` - Type of region
- `strata_included` - Strata *potentially* included in the annual index calculations
- `strata_excluded` - Strata *potentially* excluded from the annual index calculations because they have no observations of the species in the first part of the time series, see arguments `max_backcast` and `start_year`
- `index` - Strata-weighted count index (median)
- `index_q_XXX` - Strata-weighted count index (by different quantiles)
- `obs_mean` - Mean observed annual counts of birds across all routes and all years. An alternative estimate of the average relative abundance of the species in the region and year. Differences between this and the annual indices are a function of the model. For composite regions (i.e., anything other than stratum-level estimates) this average count is calculated as an area-weighted average across all strata included
- `n_routes` - Number of BBS routes that contributed data for this species, region, and year
- `n_routes_total` - Number of BBS routes that contributed data for this species and region for all years in the selected time-series, i.e., all years since `start_year`
- `n_non_zero` - Number of BBS routes on which this species was observed (i.e., count is > 0) in this region and year
- `backcast_flag` - Approximate annual average proportion of the covered species range that is free of extrapolated population trajectories. e.g., if 1.0, data cover full time-series; if 0.75, data cover 75 percent of time-series. Only calculated if `max_backcast != NULL`.

## See Also

Other indices and trends functions: [generate\\_trends\(\)](#), [plot\\_geofacet\(\)](#), [plot\\_indices\(\)](#), [plot\\_map\(\)](#)

**Examples**

```

# Using the example model for Pacific Wrens

# Generate the continental and stratum indices
i <- generate_indices(pacific_wren_model)

# Generate the continental and stratum indices using hpdi
i <- generate_indices(pacific_wren_model, hpdi = TRUE)

# Generate only country indices
i_nat <- generate_indices(pacific_wren_model, regions = "country")

# Use a custom region specification (dummy example)
library(dplyr)
ri <- bbs_strata[["bbs_cws"]]
ri <- mutate(ri, my_region = if_else(prov_state %in% "ON",
                                   "Ontario", "Rest"))

# Generate indices with these custom regions
i_custom <- generate_indices(
  pacific_wren_model,
  regions = c("country", "prov_state", "my_region"),
  regions_index = ri)

```

---

generate\_trends

*Generate regional trends*


---

**Description**

Generates trends for continent and strata and optionally for countries, states/provinces, or BCRs from analyses run on the stratifications that support these composite regions. Calculates the geometric mean annual changes in population size for composite regions.

**Usage**

```

generate_trends(
  indices,
  min_year = NULL,
  max_year = NULL,
  quantiles = c(0.025, 0.05, 0.25, 0.75, 0.95, 0.975),
  slope = FALSE,
  prob_decrease = NULL,
  prob_increase = NULL,
  hpdi = FALSE
)

```

**Arguments**

indices	List. Indices generated by generate_indices().
min_year	Numeric. Minimum year to use. Default (NULL) uses first year in data.
max_year	Numeric. Maximum year to use. Default (NULL) uses first year in data.
quantiles	Numeric vector. Quantiles to be sampled from the posterior distribution. Defaults to c(0.025, 0.05, 0.25, 0.5, 0.75, 0.95, 0.975).
slope	Logical. Whether to calculate an alternative trend metric, the slope of a log-linear regression through the annual indices. Default FALSE, which estimates the trend as the geometric mean annual rate of change between min_year and max_year. This is the end-point definition of trend that only directly incorporates information from the two years, and therefore closely tracks the annual population fluctuations in those particular years. Conceptually, this metric of trend tracks the difference between the two years. If TRUE, trend represents the slope of a linear regression through the log-transformed annual indices of abundance for all years between min_year and max_year. This definition of trend is less sensitive to the particular annual fluctuations of a given min_year and max_year. Either metric may be more or less appropriate given the user's desired inference. The appropriate choice of metric may also depend on the model and the alternate_n choice made in generate_indices. For example if the fitted model was one of the "gamye" alternatives, and the alternate_n = "nsmooth", then the default slope = FALSE option will represent the end-point difference of the smooth component, which already excludes the annual fluctuations and so has similar inferential properties as the slope = TRUE option from the "first_diff" model.
prob_decrease	Numeric vector. Percent-decrease values for which to optionally calculate the posterior probabilities (see Details). Default is NULL (not calculated). Can range from 0 to 100.
prob_increase	Numeric vector. Percent-increase values for which to optionally calculate the posterior probabilities (see Details). Default is NULL (not calculated). Can range from 0 to Inf.
hpd	Logical. Should credible intervals and limits be calculated using highest posterior density intervals instead of simple quantiles of the posterior distribution. Default is FALSE. these intervals are often a better descriptor of skewed posterior distributions, such as the predicted mean counts that the indices represent. Note hpd intervals are not stable for small percentages of the posterior distribution, and so hpd = TRUE is ignored for quantiles values between 0.33 and 0.67 (i.e., if the quantiles value defines a limit for a centered hpd interval that would include < 33% of the posterior distribution).

**Details**

The posterior probabilities can be calculated for a percent-decrease (prob\_decrease) and/or percent-increase (prob\_increase) if desired. These calculate the probability that the population has decreased/increased by at least the amount specified.

For example, a prob\_increase = 100 would result in the calculation of the probability that the population has increased by more than 100% (i.e., doubled) over the period of the trend.

Alternatively, a `prob_decrease = 50` would result in the calculation of the probability that the population has decreased by more than 50% (i.e., less than half of the population remains) over the period of the trend.

## Value

A list containing

- `trends` - data frame of calculated population trends, one row for each region in the input indices
- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata meta data
- `raw_data` - data frame of summarized counts

`trends` contains the following columns:

- `start_year` - First year of the trend
- `end_year` - Last year of the trend
- `region` - Region name
- `region_type` - Type of region
- `strata_included` - Strata *potentially* included in the annual index calculations
- `strata_excluded` - Strata *potentially* excluded from the annual index calculations because they have no observations of the species in the first part of the time series, see arguments `max_backcast` and `start_year`
- `trend` - Estimated median annual percent change over the trend time-period according to end point comparison of annual indices for the `start_year` and the `end_year`
- `trend_q_XXX` - Trend estimates by different quantiles
- `percent_change` - Median overall estimate percent change over the trend time-period
- `percent_change_q_XXX` - Percent change by different quantiles
- `slope_trend` - Estimated median annual percent change over the trend time-period, according to the slope of a linear regression through the log-transformed annual indices. (Only if `slope = TRUE`)
- `slope_trend_q_XXX` - Slope-based trend estimates by different quantiles. (Only if `slope = TRUE`)
- `width_of_95_percent_credible_interval` - Width (in percent/year) of the credible interval on the trend calculation. Calculated for the widest credible interval requested in via quantiles. Default is 95 percent CI (i.e., `trend_q_0.975 - trend_q_0.025`)
- `width_of_95_percent_credible_interval_slope` - Width (in percent/year) of the credible interval on the slope-based trend calculation. Calculated for the widest credible interval requested in via quantiles. Default is 95 percent CI (i.e., `slope_trend_q_0.975 - slope_trend_q_0.025`). (Only if `slope = TRUE`)
- `prob_decrease_XX_percent` - Proportion of the posterior distribution of `percent_change` that is below the percentage values in `prob_decrease` (if non-Null)
- `prob_increase_XX_percent` - Proportion of the posterior distribution of `percent_change` that is above the percentage values in `prob_increase` (if non-Null)

- `rel_abundance` - Mean annual index value across all years. An estimate of the average relative abundance of the species in the region. Can be interpreted as the predicted average count of the species in an average year on an average route by an average observer, for the years, routes, and observers in the existing data
- `obs_rel_abundance` - Mean observed annual count of birds across all routes and all years. An alternative estimate of the average relative abundance of the species in the region. For composite regions (i.e., anything other than stratum-level estimates) this average count is calculated as an area-weighted average across all strata included.
- `n_routes` - Number of BBS routes that contributed data for this species and region for all years in the selected time-series, i.e., all years since `start_year`
- `mean_n_routes` - Mean number of BBS routes that contributed data for this species, region, and year
- `n_strata_included` - The number of strata included in the region
- `backcast_flag` - Approximate annual average proportion of the covered species range that is free of extrapolated population trajectories. e.g., if 1.0, data cover full time-series; if 0.75, data cover 75 percent of time-series. Only calculated if `max_backcast != NULL`.

### See Also

Other indices and trends functions: [generate\\_indices\(\)](#), [plot\\_geofacet\(\)](#), [plot\\_indices\(\)](#), [plot\\_map\(\)](#)

### Examples

```
# Using the example model for Pacific Wrens...

# Generate the continental and stratum indices
i <- generate_indices(pacific_wren_model)

# Now, generate the trends
t <- generate_trends(i)

# Use the slope method
t <- generate_trends(i, slope = TRUE)

# Calculate probability of the population declining by 50%
t <- generate_trends(i, prob_decrease = 50)
```

---

get\_convergence

*Convergence metrics*

---

### Description

Calculate convergence metrics for the model run. Specifically calculates bulk and tail effective sample sizes (`ess_bulk`, `ess_tail`) and R-hat (`rhat`). Returns output very similar to `get_summary()`.

**Usage**

```
get_convergence(model_output, variables = NULL)
```

**Arguments**

`model_output` List. Model output generated by `run_model()`.

`variables` Character vector. Specific variables (e.g., "strata\_raw[1]") or variable types (e.g., "strata\_raw") for which to calculate metrics. If NULL (default) all variables are returned.

**Value**

Data frame of convergence metrics for all model variables. Contains `variable_type`, `variable`, `ess_bulk`, `ess_tail`, and `rhat`.

**See Also**

Other model assessment functions: [get\\_model\\_vars\(\)](#), [get\\_summary\(\)](#)

**Examples**

```
# Temporarily suppress convergence warning for legibility
# "The ESS has been capped to avoid unstable estimates."
opts <- options(warn = -1)

# Using the example model for Pacific Wrens

get_convergence(pacific_wren_model)
get_convergence(pacific_wren_model, variables = "strata_raw")
get_convergence(pacific_wren_model, variables = "strata_raw[9]")

# Restore warnings
options(opts)
```

---

get\_model\_vars

*Get model variables*

---

**Description**

Returns the basic model variables and/or variable types (note that most variables have different iterations for each strata and each year).

**Usage**

```
get_model_vars(model_output, all = FALSE)
```



**Arguments**

- model\_output List. Model output generated by run\_model().
- all Logical. Whether or not to return **all**, specific variables (e.g., strata\_raw[1] or just variable types (e.g., strata\_raw). Defaults to FALSE (variable types only).

**Value**

A character vector of all model variable types.

**See Also**

Other model assessment functions: [get\\_convergence\(\)](#), [get\\_summary\(\)](#)

**Examples**

```
# Using the example model for Pacific Wrens...

# List variable types
get_model_vars(pacific_wren_model)

# List all variables
get_model_vars(pacific_wren_model, all = TRUE)
```

---

get\_summary *Return the cmdstanr summary*

---

**Description**

Extract and return the model summary using cmdstanr::summary().

**Usage**

```
get_summary(model_output, variables = NULL)
```

**Arguments**

- model\_output List. Model output generated by run\_model().
- variables Character vector. Specific variables (e.g., "strata\_raw[1]") or variable types (e.g., "strata\_raw") for which to calculate metrics. If NULL (default) all variables are returned.

**Value**

A data frame of model summary statistics.

**See Also**

Other model assessment functions: [get\\_convergence\(\)](#), [get\\_model\\_vars\(\)](#)

**Examples**

```
# Temporarily suppress convergence warning for legibility
# "The ESS has been capped to avoid unstable estimates."
opts <- options(warn = -1)

# Using the example model for Pacific Wrens

get_summary(pacific_wren_model)
get_summary(pacific_wren_model, variables = "strata_raw")
get_summary(pacific_wren_model, variables = "strata_raw[9]")

# Restore warnings
options(opts)
```

---

have_bbs_data	<i>Check whether BBS data exists locally</i>
---------------	--

---

**Description**

Use this function to check if you have the BBS data downloaded and where bbsBayes2 is expecting to find it. If it returns FALSE, the data is not present; use [fetch\\_bbs\\_data\(\)](#) to retrieve it.

**Usage**

```
have_bbs_data(level = "state", release = 2023, quiet = FALSE)
```

**Arguments**

level	Character. BBS data to check, one of "all", "state", or "stop". Default "state".
release	Character/Numeric. BBS data to check, one of "all", 2020, 2022, or 2023. Default 2023.
quiet	Logical. Suppress progress messages? Default FALSE.

**Value**

TRUE if the data is found, FALSE otherwise

**See Also**

Other BBS data functions: [fetch\\_bbs\\_data\(\)](#), [load\\_bbs\\_data\(\)](#), [remove\\_cache\(\)](#)

**Examples**

```

have_bbs_data()
have_bbs_data(release = 2020)
have_bbs_data(release = "all", level = "all")

```

---

have_cmdstan	<i>Check if cmdstan is installed</i>
--------------	--------------------------------------

---

**Description**

Wrapper around `cmdstanr::cmdstan_version(error_on_NA = FALSE)` for quick check.

**Usage**

```
have_cmdstan()
```

**Details**

Used internally for skipping examples and tests if no cmdstan installed.

---

load_bbs_data	<i>Load Breeding Bird Survey data</i>
---------------	---------------------------------------

---

**Description**

Load the local, minimally processed, raw, unstratified data. The data must have been previously fetched using `fetch_bbs_data()`. This function is provided for custom explorations and is not part of the analysis workflow; `stratify()` will do the loading for you.

**Usage**

```
load_bbs_data(level = "state", release = 2023, sample = FALSE, quiet = TRUE)
```

**Arguments**

level	Character. Which type of BBS counts to use, "state" or "stop". Default "state".
release	Numeric. Which yearly release to use, 2022 (including data through 2021 field season) or 2020 (including data through 2019). Default 2022.
sample	Logical. Whether or not to use the sample data for Pacific Wrens (see <code>?bbs_data_sample</code> ). Default is FALSE. If TRUE, level and release are ignored.
quiet	Logical. Suppress progress messages? Default FALSE.

**Value**

Large list (3 elements) consisting of:

birds	Data frame of sample bird point count data per route, per year
routes	Data frame of sample yearly route data
species	Sample list of North American bird species

**See Also**

Other BBS data functions: [fetch\\_bbs\\_data\(\)](#), [have\\_bbs\\_data\(\)](#), [remove\\_cache\(\)](#)

---

load_map	<i>Load a geographic strata map</i>
----------	-------------------------------------

---

**Description**

Load one of the included spatial data files (maps) as a simple features object (sf package)

**Usage**

```
load_map(stratify_by = NULL, type = "strata")
```

**Arguments**

stratify_by	Character. Stratification type. One of "prov_state", "bcr", "latlong", "bbs_cws", "bbs_usgs".
type	Character. "strata" or political map ("North America", "Canada" or "US"/"USA"/"United States of America").

**Value**

sf polygon object

**See Also**

Other helper functions: [assign\\_prov\\_state\(\)](#), [search\\_species\(\)](#)

**Examples**

```
# Toy example with Pacific Wren sample data
# First, stratify the sample data
strat_map <- load_map(stratify_by = "bbs_cws")

# simple plot of the map
plot(strat_map)

# or use ggplot2
library(ggplot2)
```

```
ggplot(data = strat_map) +  
  geom_sf(aes(fill = strata_name), show.legend = FALSE)
```

---

pacific\_wren\_model      *Example model output*

---

## Description

Example model output from running a hierarchical first difference model on the included sample data for Pacific Wrens.

## Usage

```
pacific_wren_model
```

## Format

```
pacific_wren_model:
```

A list output from `run_model()` with 4 items

- `model_fit` - cmdstanr model output
- `model_data` - list of data formatted for use in Stan modelling
- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata meta data
- `raw_data` - data frame of summarized counts

## Examples

```
# Code to replicate:  
## Not run:  
pacific_wren_model <- stratify(by = "bbs_cws", sample_data = TRUE) %>%  
  prepare_data() %>%  
  prepare_model(model = "first_diff", set_seed = 111) %>%  
  run_model(chains = 2, iter_sampling = 20, iter_warmup = 20, set_seed = 111)  
  
## End(Not run)
```

---

plot\_geofacet

*Create geofacet plot of population trajectories by province/state*


---

### Description

Generate a faceted plot of population trajectories by province/state. Only possible if indices created by `generate_indices()` include the `prov_state` region. All geofacet plots have one facet per state/province, so if there are multiple strata per facet, these can be plotted as separate trajectories within each facet (`multiple = TRUE`).

### Usage

```
plot_geofacet(
  indices,
  ci_width = 0.95,
  multiple = FALSE,
  trends = NULL,
  slope = FALSE,
  add_observed_means = FALSE,
  col_viridis = FALSE
)
```

### Arguments

<code>indices</code>	List. Indices generated by <code>generate_indices()</code> .
<code>ci_width</code>	Numeric. Quantile defining the width of the plotted credible interval. Defaults to 0.95 (lower = 0.025 and upper = 0.975). Note these quantiles need to have been precalculated in <code>generate_indices()</code> .
<code>multiple</code>	Logical. Whether to plot multiple strata-level trajectories within each prov/state facet. Default FALSE.
<code>trends</code>	List. (Optional) Output generated by <code>generate_trends()</code> . If included trajectories are coloured based on the same colour scale used in <code>plot_map</code> .
<code>slope</code>	Logical. If trends included, whether colours in the plot should be based on slope trends. Default FALSE.
<code>add_observed_means</code>	Logical. Whether to include points indicating the observed mean counts. Default FALSE. Note: scale of observed means and annual indices may not match due to imbalanced sampling among routes.
<code>col_viridis</code>	Logical. Should the colour-blind-friendly "viridis" palette be used. Default FALSE.

### Value

ggplot object

**See Also**

Other indices and trends functions: [generate\\_indices\(\)](#), [generate\\_trends\(\)](#), [plot\\_indices\(\)](#), [plot\\_map\(\)](#)

**Examples**

```
# Using the example model for Pacific Wrens...

# Generate indices
i <- generate_indices(pacific_wren_model,
                     regions = c("stratum", "prov_state"))

# Generate trends
t <- generate_trends(i)

# Now make the geofacet plot.
plot_geofacet(i, trends = t, multiple = TRUE)
plot_geofacet(i, trends = t, multiple = TRUE, col_viridis = TRUE)
plot_geofacet(i, multiple = TRUE)
plot_geofacet(i, trends = t, multiple = FALSE)
plot_geofacet(i, multiple = FALSE)

# With different ci_width, specify desired quantiles in indices
i <- generate_indices(pacific_wren_model,
                     regions = c("stratum", "prov_state"),
                     quantiles = c(0.005, 0.995))

plot_geofacet(i, multiple = FALSE, ci_width = 0.99)
```

---

plot\_indices

*Generate plots of index trajectories by stratum*

---

**Description**

Generates the indices plot for each stratum modelled.

**Usage**

```
plot_indices(
  indices = NULL,
  ci_width = 0.95,
  min_year = NULL,
  max_year = NULL,
  title = TRUE,
  title_size = 20,
  axis_title_size = 18,
  axis_text_size = 16,
  line_width = 1,
```

```

    add_observed_means = FALSE,
    add_number_routes = FALSE
  )

```

### Arguments

<code>indices</code>	List. Indices generated by <code>generate_indices()</code> .
<code>ci_width</code>	Numeric. Quantile defining the width of the plotted credible interval. Defaults to 0.95 (lower = 0.025 and upper = 0.975). Note these quantiles need to have been precalculated in <code>generate_indices()</code> .
<code>min_year</code>	Numeric. Minimum year to plot.
<code>max_year</code>	Numeric. Maximum year to plot.
<code>title</code>	Logical. Whether to include a title on the plot.
<code>title_size</code>	Numeric. Font size of plot title. Defaults to 20
<code>axis_title_size</code>	Numeric. Font size of axis titles. Defaults to 18
<code>axis_text_size</code>	Numeric. Font size of axis text. Defaults to 16
<code>line_width</code>	Numeric. Size of the trajectory line. Defaults to 1
<code>add_observed_means</code>	Logical. Whether to include points indicating the observed mean counts. Default FALSE. Note: scale of observed means and annual indices may not match due to imbalanced sampling among routes.
<code>add_number_routes</code>	Logical. Whether to superimpose dotplot showing the number of BBS routes included in each year. This is useful as a visual check on the relative data-density through time because in most cases the number of observations increases over time.

### Value

List of ggplot2 plots, each item being a plot of a stratum's indices.

### See Also

Other indices and trends functions: [generate\\_indices\(\)](#), [generate\\_trends\(\)](#), [plot\\_geofacet\(\)](#), [plot\\_map\(\)](#)

### Examples

```

# Using the example model for Pacific Wrens...

# Generate country, continent, and stratum indices
i <- generate_indices(model_output = pacific_wren_model,
                     regions = c("country", "continent", "stratum"))

# Now, plot_indices() will generate a list of plots for all regions
plots <- plot_indices(i)

```



```

# To view any plot, use [[i]]
plots[[1]]

names(plots)

# Suppose we wanted to access the continental plot. We could do so with
plots[["continent"]]

# You can specify to only plot a subset of years using min_year and max_year

# Plots indices from 2015 onward
p_2015_min <- plot_indices(i, min_year = 2015)
p_2015_min[["continent"]]

#Plot up indices up to the year 2017
p_2017_max <- plot_indices(i, max_year = 2017)
p_2017_max[["continent"]]

#Plot indices between 2011 and 2016
p_2011_2016 <- plot_indices(i, min_year = 2011, max_year = 2016)
p_2011_2016[["continent"]]

```

---

plot\_map

*Generate a map of trends by strata*


---

## Description

plot\_map() allows you to generate a colour-coded map of the percent change in species trends for each strata.

## Usage

```

plot_map(
  trends,
  slope = FALSE,
  title = TRUE,
  alternate_column = NULL,
  col_viridis = FALSE,
  strata_custom = NULL
)

```

## Arguments

trends	List. Trends generated by generate_trends().
slope	Logical. Whether or not to map values of the alternative trend metric (slope of a log-linear regression) if slope = TRUE was used in generate_trends(), through the annual indices. Default FALSE.

title	Logical. Whether or not to include a title with species. Default TRUE.
alternate_column	Character, Optional name of numerical column in trends dataframe to plot. If one of the columns with "trend" in the title, (e.g., trend_q_0.05 then the colour scheme and breaks will match those used in the default trend maps)
col_viridis	Logical. Should the colour-blind-friendly "viridis" palette be used. Default FALSE.
strata_custom	(sf) Data Frame. Data frame of modified existing stratification, or a sf spatial data frame with polygons defining the custom stratifications. See details on strata_custom in stratify().

### Value

a ggplot2 plot

### See Also

Other indices and trends functions: [generate\\_indices\(\)](#), [generate\\_trends\(\)](#), [plot\\_geofacet\(\)](#), [plot\\_indices\(\)](#)

### Examples

```
# Using the example model for Pacific Wrens...

# Generate the continental and stratum indices
i <- generate_indices(pacific_wren_model)

# Now generate trends
t <- generate_trends(i, slope = TRUE)

# Generate the map (without slope trends)
plot_map(t)

# Generate the map (with slope trends)
plot_map(t, slope = TRUE)

# Viridis
plot_map(t, col_viridis = TRUE)

# Generate a map (with alternate column - lower 95% Credible limit)
plot_map(t, alternate_column = "trend_q_0.05")
```

**Description**

Check and filter the stratified data by minimum required samples for modelling, and prepare data format for use by models.

**Usage**

```
prepare_data(
  strata_data,
  min_year = NULL,
  max_year = NULL,
  min_n_routes = 3,
  min_max_route_years = 3,
  min_mean_route_years = 1,
  quiet = FALSE
)
```

**Arguments**

strata_data	List. Stratified data generated by <code>stratify()</code>
min_year	Numeric. Minimum year to use. Default (NULL) uses first year in data.
max_year	Numeric. Maximum year to use. Default (NULL) uses first year in data.
min_n_routes	Numeric. Required minimum routes per strata where species has been observed. Default 3.
min_max_route_years	Numeric. Required minimum number of years with non-zero observations of species on at least 1 route. Default 3. Only retain strata with at least one route where the species was observed at least once in this many years.
min_mean_route_years	Numeric. Required minimum average of years per route with the species observed. Default 1. Only retain strata where the average number of years the species was observed per route is greater than this value.
quiet	Logical. Suppress progress messages? Default FALSE.

**Value**

List of prepared (meta) data to be used for modelling and further steps.

- `model_data` - list of data formatted for use in Stan modelling
- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata meta data
- `raw_data` - data frame of summarized counts used to create `model_data` (just formatted more nicely)

**See Also**

Other Data prep functions: [prepare\\_model\(\)](#), [prepare\\_spatial\(\)](#), [stratify\(\)](#)

**Examples**

```
# Toy example with Pacific Wren sample data

# First, stratify the sample data

s <- stratify(by = "bbs_cws", sample_data = TRUE)

# Prepare the stratified data for use in a model. In this
# toy example, we will set the minimum year as 2009 and
# maximum year as 2018, effectively only setting up to
# model 10 years of data.

p <- prepare_data(s, min_year = 2009, max_year = 2018)
```

---

```
prepare_model
```

```
Prepare model parameters
```

---

**Description**

Calculate and format the prepared data for use in modelling. Different parameters are defined for different types of models (see `?bbs_models` for a list of models included in `bbsBayes2`).

**Usage**

```
prepare_model(
  prepared_data,
  model,
  model_variant = "hier",
  model_file = NULL,
  use_pois = FALSE,
  heavy_tailed = TRUE,
  n_knots = NULL,
  basis = "mgcv",
  calculate_nu = FALSE,
  calculate_log_lik = FALSE,
  calculate_cv = FALSE,
  cv_k = 10,
  cv_fold_groups = "obs_n",
  cv_omit_singles = TRUE,
  set_seed = NULL,
  quiet = FALSE
)
```

**Arguments**

`prepared_data` List. Prepared data generated by `prepare_data()` (if `model_variant` is not `spatial`) or `prepare_spatial()` (if `model_variant` is `"spatial"`).

model	Character. Type of model to use, must be one of "first_diff" (First Differences), "gam" (General Additive Model), "gamy" (General Additive Model with Year Effect), or "slope" (Slope model).
model_variant	Character. Model variant to use, must be one of "nonhier" (Non-hierarchical), "hier" (Hierarchical; default), or "spatial" (Spatially explicit).
model_file	Character. Optional location of a custom Stan model file to use.
use_pois	Logical. Whether to use an Over-Dispersed Poisson model (TRUE) or an Negative Binomial model (FALSE; default).
heavy_tailed	Logical. Whether extra-Poisson error distributions should be modelled as a t-distribution, with heavier tails than the standard normal distribution. Default TRUE. Recent results suggest this is best even though it requires much longer convergence times. Can only be set to FALSE with Poisson models (i.e. use_pois = TRUE).
n_knots	Numeric. Number of knots for "gam" and "gamy" models
basis	Character. Basis function to use for GAM smooth, one of "original" or "mgcv". Default is "original", the same basis used in Smith and Edwards 2020. "mgcv" is an alternate that uses the "tp" basis from the package mgcv (also used in brms, and rstanarm). If using the "mgcv" option, the user may want to consider adjusting the prior distributions for the parameters and their precision.
calculate_nu	Logical. Whether to calculate the nu parameter as a factor of $\gamma(2, 0.1)$ . Default FALSE.
calculate_log_lik	Logical. Whether to calculate point-wise log-likelihood of the data given the model. Default FALSE.
calculate_cv	Logical. Whether to use bbsBayes2' cross validation. Note this is <b>experimental</b> . See Details.
cv_k	Numeric. The number of K folds to include (only relevant if calculate_cv = TRUE). Default 10. Note this is <b>experimental</b> .
cv_fold_groups	Character. The data column to use when determining the grouping level of the observations to be assigned to different fold groups. Must be one of obs_n (default) or routes (only relevant if calculate_cv = TRUE). Note this is <b>experimental</b> . See the <a href="#">models article</a> for more details.
cv_omit_singles	Logical. Whether to omit test groups with no replication (only relevant if calculate_cv = TRUE). Default TRUE. See the <a href="#">models article</a> for more details.
set_seed	Numeric. If NULL (default) no seed is set. Otherwise an integer number to be used with <code>withr::with_seed()</code> internally to ensure reproducibility.
quiet	Logical. Suppress progress messages? Default FALSE.

## Details

There are two ways you can customize the model run. The first is to supply a custom `model_file` created with the `copy_model_file()` function and then edited by hand.

Second, you can edit or overwrite the initialization parameters (`init_values`) in the output of `prepare_model()` to customize the `init` supplied to `cmdstanr::sample()`. You can supply these parameters in anyway that `cmdstanr::sample()` accepts the `init` argument.

To implement bbsBayes2' version of cross validation, set `calculate_cv = TRUE`. You can set up your own system for cross validation by modifying the `fold`s list-item in the output of `prepare_model()`.

**Note this is considered experimental.**

See the [models article](#) for more advanced examples and explanations.

### Value

A list of prepared data.

- `model_data` - list of data formatted for use in Stan modelling
- `init_values` - list of initialization parameters
- `fold`s - a vector of k-fold groups each observation is assigned to (if `calculate_cv = TRUE`), or NULL
- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata meta data
- `raw_data` - data frame of summarized counts used to create `model_data` (just formatted more nicely)

### See Also

Other Data prep functions: [prepare\\_data\(\)](#), [prepare\\_spatial\(\)](#), [stratify\(\)](#)

### Examples

```
s <- stratify(by = "bbs_cws", sample_data = TRUE)
p <- prepare_data(s)
pm <- prepare_model(p, model = "first_diff", model_variant = "hier")
```

---

```
prepare_spatial
```

---

```
Define neighbouring strata for spatial analyses
```

---

### Description

Given the prepared data and a spatial data frame of polygons outlining strata, identify a neighbourhood matrix for use in modelling.

### Usage

```
prepare_spatial(
  prepared_data,
  strata_map,
  voronoi = FALSE,
  nearest_fill = FALSE,
  island_link_dist_factor = 1.2,
  buffer_type = "buffer",
  buffer_dist = 10000,
```

```

    add_map = NULL,
    label_size = 3,
    quiet = FALSE
  )

```

### Arguments

prepared_data	List. Prepared data generated by prepare_data().
strata_map	sf Data Frame. sf map of the strata in (MULTI)POLYGONS. Must have column "strata_name" matching strata output from prepare_data().
voronoi	Logical. Whether or not to use Voroni method. Default FALSE.
nearest_fill	Logical. For strata with no neighbours, whether or not to fill in by centroids of the 2 nearest neighbours when <b>not</b> using the Voronoi method. Default FALSE.
island_link_dist_factor	Numeric. Distances within a factor of this amount are considered nearest strata neighbours. Used when linking otherwise isolated islands of strata, when <b>not</b> using the Voronoi method. Default 1.2.
buffer_type	Character. Which buffer type to use when using the Voronoi method. Must be one of buffer (default) or convex_hull. See Details for specifics.
buffer_dist	Numeric. Distance to buffer and link the strata if not connected when using the Voronoi method. Units are that of sf::st_crs(strata_map). This is the <i>starting</i> distance if buffer_type = "buffer" or the final distance if buffer_type = "convex_hull". Default 10000. See Details.
add_map	sf object. Spatial data to add to map output.
label_size	Numeric. Size of the labels on the map. For data with many different strata it can be useful to reduce the size of the labels. Default 3.
quiet	Logical. Suppress progress messages? Default FALSE.

### Details

When using the Voronoi method, a buffer is used to fill around and link strata together. If the buffer\_type is buffer, buffer\_dist is the starting distance over which to buffer. If not all strata are linked, this distance is increased by 10% and applied again, repeating until all strata are linked. If buffer\_type is convex\_hull, then a convex hull is used to link up the strata before applying a buffer at a distance of buffer\_dist. Note that all distances are in the units of sf::st\_crs(strata\_map).

### Value

List of prepared (meta) data to be used for modelling and further steps.

- spatial\_data - list of samples, nodes, adjacent matrix and map visualizing the matrix
- model\_data - list of data formatted for use in Stan modelling
- meta\_data - meta data defining the analysis
- meta\_strata - data frame listing strata meta data
- raw\_data - data frame of summarized counts used to create model\_data (just formatted more nicely)

**See Also**

Other Data prep functions: [prepare\\_data\(\)](#), [prepare\\_model\(\)](#), [stratify\(\)](#)

**Examples**

```
map <- load_map("bbs_cws")
s <- stratify(by = "bbs_cws", sample_data = TRUE)
p <- prepare_data(s, min_max_route_years = 2)
sp <- prepare_spatial(p, map)

# Visually explore the spatial linkages
sp$map

# Overlay subset strata map on original mapping data
sp <- prepare_spatial(p, map, add_map = map)
sp$map
```

---

remove\_cache

*Remove bbsBayes2 cache*

---

**Description**

Remove all or some of the data downloaded via `fetch_bbs_data()` as well as model executables created by `cmdstanr::cmdstan_model()` via `run_model()`.

**Usage**

```
remove_cache(type = "bbs_data", level = "all", release = "all")
```

**Arguments**

type	Character. Which cached data to remove. One of "all", "bbs_data", or "models". If "all", removes entire cache directory (and all data contained therein). If "bbs_data", removes only BBS data downloaded with <code>fetch_bbs_data()</code> . If "models", removes only model executables compiled when <code>run_models()</code> is run.
level	Character. BBS data to remove, one of "all", "state", or "stop". Only applies if <code>type = "bbs_data"</code> . Default "all".
release	Character/Numeric. BBS data to remove, one of "all", 2020, 2022, or 2023. Only applies if <code>type = "bbs_data"</code> . Default "all".

**Value**

Nothing



**See Also**

Other BBS data functions: [fetch\\_bbs\\_data\(\)](#), [have\\_bbs\\_data\(\)](#), [load\\_bbs\\_data\(\)](#)

**Examples**

```
## Not run:
# Remove everything
remove_cache(type = "all")

# Remove all BBS data files (but not the dir)
remove_cache(level = "all", release = "all")

# Remove all 'stop' data
remove_cache(level = "stop", release = "all")

# Remove all 2020 data
remove_cache(level = "all", release = 2020)

# Remove 2020 stop data
remove_cache(level = "stop", release = 2020)

# Remove all model executables
remove_cache(type = "model")

## End(Not run)
```

---

run\_model

*Run Bayesian model*

---

**Description**

Run Bayesian model with `cmdstandr::sample()` using prepare data and model parameters specified in previous steps.

**Usage**

```
run_model(
  model_data,
  refresh = 100,
  chains = 4,
  parallel_chains = 4,
  iter_warmup = 1000,
  iter_sampling = 1000,
  adapt_delta = 0.8,
  max_treedepth = 11,
  k = NULL,
  output_basename = NULL,
```

```

output_dir = ".",
save_model = TRUE,
overwrite = FALSE,
retain_csv = FALSE,
set_seed = NULL,
quiet = FALSE,
...
)

```

## Arguments

model_data	List. Model data generated by <code>prepare_model()</code> .
refresh	Numeric. Passed to <code>cmdstanr::sample()</code> . Number of iterations between screen updates. If 0, only errors are shown.
chains	Numeric. Passed to <code>cmdstanr::sample()</code> . Number of Markov chains to run.
parallel_chains	Numeric. Passed to <code>cmdstanr::sample()</code> . Maximum number of chains to run in parallel.
iter_warmup	Numeric. Passed to <code>cmdstanr::sample()</code> . Number of warmup iterations per chain.
iter_sampling	Numeric. Passed to <code>cmdstanr::sample()</code> . Number of sampling (post-warmup) iterations per chain.
adapt_delta	Numeric. Passed to <code>cmdstanr::sample()</code> . The adaptation target acceptance statistic.
max_treedepth	Numeric. Passed to <code>cmdstanr::sample()</code> . The maximum allowed tree depth for the NUTS engine. See <code>?cmdstanr::sample</code> .
k	Numeric. The K-fold group to run for cross-validation. Only relevant if folds defined by <code>prepare_model(estimate_cv = TRUE)</code> or custom definition. See <code>?prepare_model</code> or the <a href="#">models article</a> for more details.
output_basename	Character. Name of the files created as part of the Stan model run and the final model output RDS file if <code>save_model = TRUE</code> . Defaults to a character string that is unique to the species, model, model_variant, and system time of <code>model_run()</code> call (nearest minute).
output_dir	Character. Directory in which all model files will be created. Defaults to the working directory, but recommend that the user sets this to a particular existing directory for better file organization.
save_model	Logical. Whether or not to save the model output to file as an RDS object with all required data. Defaults to TRUE.
overwrite	Logical. Whether to overwrite an existing model output file when saving.
retain_csv	Logical. Whether to retain the Stan csv files after the model has finished running and the fitted object has been saved. Defaults to FALSE because csv files duplicate information saved in the model output file save object, when <code>save_model = TRUE</code> , and so for file organization and efficient use of memory, these are deleted by default.

set_seed	Numeric. If NULL (default) no seed is set. Otherwise an integer number to be used with <code>withr::with_seed()</code> internally to ensure reproducibility.
quiet	Logical. Suppress progress messages? Default FALSE.
...	Other arguments passed on to <code>cmdstanr::sample()</code> .

### Details

The model is set up in `prepare_model()`. The `run_model()` function does the final (and often long-running) step of actually running the model. Here is where you can tweak how the model will be run (iterations etc.).

See the [models article](#) for more advanced examples and explanations.

### Value

List model fit and other (meta) data.

- `model_fit` - cmdstanr model output
- `model_data` - list of data formatted for use in Stan modelling
- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata meta data
- `raw_data` - data frame of summarized counts

### See Also

Other modelling functions: [copy\\_model\\_file\(\)](#), [save\\_model\\_run\(\)](#)

### Examples

```
s <- stratify(by = "bbs_cws", sample_data = TRUE)
p <- prepare_data(s)
pm <- prepare_model(p, model = "first_diff", model_variant = "hier")

# Run model (quick and dirty)
m <- run_model(pm, iter_warmup = 20, iter_sampling = 20, chains = 2)
```

---

save_model_run	<i>Save output of run_model()</i>
----------------	-----------------------------------

---

### Description

This function closely imitates `cmdstanr::save_object()` but saves the entire model output object from `run_model()` which contains more details regarding data preparation (stratification etc.).

### Usage

```
save_model_run(model_output, retain_csv = TRUE, path = NULL, quiet = FALSE)
```

**Arguments**

model_output	List. Model output generated by run_model().
retain_csv	Logical Should the Stan csv files be retained. Defaults to TRUE if user calls function directly. However, when this function is called internally by run_model this is set to FALSE.
path	Character. Optional file path to use for saved data. Defaults to the file path used for the original run.
quiet	Logical. Suppress progress messages? Default FALSE.

**Details**

Files are saved to path, or if not provided, to the original location of the Stan model run files (if the original files exist).

**Value**

Nothing. Creates an .rds file at path.

**See Also**

Other modelling functions: [copy\\_model\\_file\(\)](#), [run\\_model\(\)](#)

**Examples**

```
# By default, the model is saved as an RDS file during `run_model()`

# But you can also deliberately save the file (here with an example model)
save_model_run(pacific_wren_model, path = "my_model.rds")

# Clean up
unlink("my_model.rds")
```

---

search_species	<i>Search for species</i>
----------------	---------------------------

---

**Description**

A helper function for finding the appropriate species name for use in stratify().

**Usage**

```
search_species(species, combine_species_forms = TRUE)
```

**Arguments**

`species` Character/Numeric. Search term, either name in English or French, AOU code, or scientific genus or species. Matches by regular expression but ignores case.

`combine_species_forms` Logical. Whether or not to search the combined species data or the uncombined species. Note that this results in different species names.

**Value**

Subset of the BBS species data frame matching the species pattern.

**See Also**

Other helper functions: [assign\\_prov\\_state\(\)](#), [load\\_map\(\)](#)

**Examples**

```
# Search for various terms
search_species("Paridae")
search_species("chickadee")
search_species("mésang")
search_species("Poecile")
search_species(7360)
search_species(73)
search_species("^73") # Use regex to match aou codes starting with 73
search_species("blue grouse")
search_species("sooty grouse")

# To combine or not
search_species("blue grouse", combine_species_forms = FALSE)
search_species("sooty grouse", combine_species_forms = FALSE)
search_species("northern flicker")
search_species("northern flicker", combine_species_forms = FALSE)
```

---

species\_forms

*Species forms*

---

**Description**

Species forms which will be combined if `combine_species_forms` is TRUE in `stratify()`.

**Usage**

```
species_forms
```

**Format**

species\_forms:

A data frame with 13 rows and 5 columns

- aou\_unid - The AOU id number which will identify the combined unidentified form
- ensligh\_original - The English name of the original 'unidentified' form
- english\_combined - The English name of the new 'combined' forms
- french\_combined - The French name of the new 'combined' forms
- aou\_id - The AOU id numbers of all the forms which will be combined

**Examples**

```
species_forms
```

---

```
stratify
```

*Stratify and filter Breeding Bird Survey data*

---

**Description**

Assign count data to strata and filter by species of interest. Routes are assigned to strata based on their geographic location and the stratification specified by the user. Species are filtered by matching English, French or Scientific names to those in the BBS species data (see `search_species()` for a flexible search to identify correct species names).

**Usage**

```
stratify(
  by,
  species,
  strata_custom = NULL,
  combine_species_forms = TRUE,
  release = 2023,
  sample_data = FALSE,
  return_omitted = FALSE,
  quiet = FALSE
)
```

**Arguments**

<code>by</code>	Character. Stratification type. Either an established type, one of "prov_state", "bcr", "latlong", "bbs_cws", "bbs_usgs", or a custom name (see <code>strata_custom</code> for details).
<code>species</code>	Character. Bird species of interest. Can be specified by English, French, or Scientific names, or AOU code. Use <code>search_species()</code> for loose matching to find the exact name/code needed.
<code>strata_custom</code>	(sf) Data Frame. Data frame of modified existing stratification, or a sf spatial data frame with polygons defining the custom stratifications. See Details.

combine_species_forms	Logical. Whether to combine ambiguous species forms. Default TRUE. See Details.
release	Numeric. Which yearly release to use, 2022 (including data through 2021 field season) or 2020 (including data through 2019). Default 2022.
sample_data	Logical. Use sample data (just Pacific Wrens). Default FALSE.
return_omitted	Logical. Whether or not to return a data frame of route-years which were omitted during stratification as they did not overlap with any stratum. For checking and troubleshooting. Default FALSE.
quiet	Logical. Suppress progress messages? Default FALSE.

### Details

To define a custom subset of an existing stratification, specify the stratification in by (e.g., "bbs\_cws") and then supply a subset of `bbs_strata[["bbs_cws"]]` to `strata_custom` (see examples).

To define a completely new custom stratification, specify the name you would like use in by (e.g., "east\_west\_divide") and then supply a spatial data frame with polygons identifying the different strata to `strata_custom`. Note that this data must have a column called `strata_name` which names all the strata contained (see examples).

If `combine_species_forms` is TRUE (default), species with multiple forms (e.g., "unid. Dusky Grouse / Sooty Grouse") are included in overall species groupings (i.e., "unid." are combined with "Dusky Grouse" and "Sooty Grouse" into "Blue Grouse (Dusky/Sooty)"). If the user wishes to keep the forms separate, `combine_species_forms` can be set to FALSE. See the data frame `species_forms`, for which species are set to be combined with which other species.

See `vignette("stratification", package = "bbsBayes2")` and the article [custom stratification](#) for more details.

### Value

List of (meta) data.

- `meta_data` - meta data defining the analysis
- `meta_strata` - data frame listing strata names and area for all strata relevant to the data (i.e. some may have been removed due to lack of count data). Contains at least `strata_name` (the label of the stratum), and `area_sq_km` (area of the stratum).
- `birds_strata` - data frame of stratified count-level data filtered by species
- `routes_strata` - data frame of stratified route-level data filtered by species

### See Also

Other Data prep functions: [prepare\\_data\(\)](#), [prepare\\_model\(\)](#), [prepare\\_spatial\(\)](#)

### Examples

```
# Sample Data - USGS BBS strata
s <- stratify(by = "bbs_usgs", sample_data = TRUE)
```

```

# Full data - species and stratification
# Use `search_species()` to get correct species name

# Stratify by CWS BBS strata
s <- stratify(by = "bbs_cws", species = "Common Loon")

# Use use English, French, Scientific, or AOU codes for species names
s <- stratify(by = "bbs_cws", species = "Plongeon huard")
s <- stratify(by = "bbs_cws", species = 70)
s <- stratify(by = "bbs_cws", species = "Gavia immer")

# Stratify by Bird Conservation Regions
s <- stratify(by = "bcr", species = "Great Horned Owl")

# Stratify by CWS BBS strata
s <- stratify(by = "bbs_cws", species = "Canada Jay")

# Stratify by State/Province/Territory only
s <- stratify(by = "prov_state", species = "Common Loon")
s <- stratify(by = "prov_state", species = "Plongeon huard")
s <- stratify(by = "prov_state", species = 70)

# Stratify by blocks of 1 degree of latitude X 1 degree of longitude
s <- stratify(by = "latlong", species = "Snowy Owl")

# Check routes omitted by stratification
s <- stratify(by = "latlong", species = "Snowy Owl", return_omitted = TRUE)
s[["routes_omitted"]]

# Use combined or non-combined species forms

search_species("Sooty grouse")
s <- stratify(by = "bbs_usgs", species = "Blue Grouse (Dusky/Sooty)")
nrow(s$birds_strata) # Contains all Dusky, Sooty and unidentified

search_species("Sooty grouse", combine_species_forms = FALSE)
s <- stratify(by = "bbs_usgs", species = "unid. Dusky Grouse / Sooty Grouse",
              combine_species_forms = FALSE)
nrow(s$birds_strata) # Contains *only* unidentified

# Stratify by a subset of an existing stratification
library(dplyr)
my_cws <- filter(bbs_strata[["bbs_cws"]], country_code == "CA")
s <- stratify(by = "bbs_cws", strata_custom = my_cws, species = "Snowy Owl")

my_bcr <- filter(bbs_strata[["bcr"]], strata_name == "BCR8")
s <- stratify(by = "bcr", strata_custom = my_bcr,
              species = "Yellow-rumped Warbler (all forms)")

# Stratify by Custom stratification, using sf map object

```



```
# e.g. with WBPHS stratum boundaries 2019
# available: https://ecos.fws.gov/ServCat/Reference/Profile/142628

## Not run:
map <- sf::read_sf("../WBPHS_Stratum_Boundaries_2019") %>%
  rename(strata_name = STRAT) # stratify expects this column

s <- stratify(by = "WBPHS_2019", strata_map = map)

## End(Not run)
```

# Index

## \* BBS data functions

fetch\_bbs\_data, 8  
have\_bbs\_data, 18  
load\_bbs\_data, 19  
remove\_cache, 32

## \* Data prep functions

prepare\_data, 26  
prepare\_model, 28  
prepare\_spatial, 30  
stratify, 38

## \* datasets

bbs\_data\_sample, 5  
bbs\_models, 5  
bbs\_strata, 6  
pacific\_wren\_model, 21  
species\_forms, 37

## \* helper functions

assign\_prov\_state, 3  
load\_map, 20  
search\_species, 36

## \* indices and trends functions

generate\_indices, 9  
generate\_trends, 12  
plot\_geofacet, 22  
plot\_indices, 23  
plot\_map, 25

## \* model assessment functions

get\_convergence, 15  
get\_model\_vars, 16  
get\_summary, 17

## \* modelling functions

copy\_model\_file, 7  
run\_model, 33  
save\_model\_run, 35

assign\_prov\_state, 3, 20, 37

bbs\_data\_sample, 5  
bbs\_models, 5  
bbs\_strata, 6

bbsBayes2-defunct, 4, 4  
bbsBayes2-deprecated, 4, 4

copy\_model\_file, 7, 35, 36

fetch\_bbs\_data, 8, 18, 20, 33

generate\_indices, 9, 15, 23, 24, 26  
generate\_trends, 11, 12, 23, 24, 26  
get\_convergence, 15, 17, 18  
get\_model\_vars, 16, 16, 18  
get\_summary, 16, 17, 17

have\_bbs\_data, 9, 18, 20, 33  
have\_cmdstan, 19

load\_bbs\_data, 9, 18, 19, 33  
load\_map, 3, 20, 37

pacific\_wren\_model, 21  
plot\_geofacet, 11, 15, 22, 24, 26  
plot\_indices, 11, 15, 23, 23, 26  
plot\_map, 11, 15, 23, 24, 25  
prepare\_data, 26, 30, 32, 39  
prepare\_model, 27, 28, 32, 39  
prepare\_spatial, 27, 30, 30, 39

remove\_cache, 9, 18, 20, 32  
run\_model, 7, 33, 36

save\_model\_run, 7, 35, 35  
search\_species, 3, 20, 36  
species\_forms, 37  
stratify, 27, 30, 32, 38