

# Package ‘S4DM’

May 11, 2026

**Title** Small Sample Size Species Distribution Modeling

**Version** 0.0.2

## Description

Implements a set of distribution modeling methods that are suited to species with small sample sizes (e.g., poorly sampled species or rare species). While these methods can also be used on well-sampled taxa, they are united by the fact that they can be utilized with relatively few data points. More details on the currently implemented methodologies can be found in Maitner et al. (2026) <[doi:10.1002/ecog.08112](https://doi.org/10.1002/ecog.08112)>, Drake and Richards (2018) <[doi:10.1002/ecs2.2373](https://doi.org/10.1002/ecs2.2373)>, Drake (2015) <[doi:10.1002.1](https://doi.org/10.1002/2020.1)>.

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**License** MIT + file LICENSE

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ensemble_range_map	<i>Generate ensemble predictions from S4DM range maps</i>
--------------------	---

---

## Description

This function evaluates model quality and creates an ensemble of the model outputs. This function uses 5-fold, spatially stratified, cross-validation to evaluate distribution model quality. Ensembles of algorithms spanning the sensitivity-specificity gradient can improve predictions for poorly sampled species (Maitner et al. 2026).

## Usage

```
ensemble_range_map(
  occurrences,
  env,
  method = NULL,
  presence_method = NULL,
  background_method = NULL,
  bootstrap = "none",
  bootstrap_reps = 100,
  quantile = 0.05,
  constraint_regions = NULL,
  background_buffer_width = NULL,
  ...
)
```

**Arguments**

occurrences	Presence coordinates in long,lat format.
env	Environmental SpatRaster(s)
method	Optional. If supplied, both presence and background density estimation will use this method.
presence_method	Optional. Method for estimation of presence density.
background_method	Optional. Method for estimation of background density.
bootstrap	Character. One of "none" (the default, no bootstrapping), "numbag" (presence function is bootstrapped), or "doublebag" (presence and background functions are bootstrapped).
bootstrap_reps	Integer. Number of bootstrap replicates to use (default is 100)
quantile	Quantile to use for thresholding. Default is 0.05 (5 pct training presence). Set to 0 for minimum training presence (MTP).
constraint_regions	See get_env_bg documentation
background_buffer_width	Numeric or NULL. Width (meters or map units) of buffer to use to select background environment. If NULL, uses max dist between nearest occurrences.
...	Additional parameters passed to internal functions.

**Details**

Current plug-and-play methods include: "gaussian", "kde", "vine", "rangebagging", "lobagoc", and "none". Current density ratio methods include: "ulsif", "rulsif".

**Value**

List object containing elements (1) spatRaster ensemble layer showing the proportion of maps that are included in the range across the ensemble, (2) spatRasters for individual models, and (3) model quality information.

**Note**

Either method or both presence\_method and background\_method must be supplied.

**Examples**

```
# load in sample data

library(S4DM)
library(terra)

# occurrence points
```

```
data("sample_points")
occurrences <- sample_points

# environmental data
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))

# rescale the environmental data

env <- scale(env)

ensemble <- ensemble_range_map(occurrences = occurrences,
                               env = env,
                               method = NULL,
                               presence_method = c("gaussian", "kde"),
                               background_method = "gaussian",
                               quantile = 0.05,
                               background_buffer_width = 100000 )
```

---

evaluate\_range\_map      *Evaluate S4DM range map quality*

---

## Description

This function uses 5-fold, spatially stratified, cross-validation to evaluate distribution model quality.

## Usage

```
evaluate_range_map(
  occurrences,
  env,
  method = NULL,
  presence_method = NULL,
  background_method = NULL,
  bootstrap = "none",
  bootstrap_reps = 100,
  quantile = 0.05,
  constraint_regions = NULL,
  background_buffer_width = NULL,
  standardize_preds = TRUE,
  ...
)
```

## Arguments

occurrences	Presence coordinates in long,lat format.
env	Environmental SpatRaster(s)

method	Optional. If supplied, both presence and background density estimation will use this method.
presence_method	Optional. Method for estimation of presence density.
background_method	Optional. Method for estimation of background density.
bootstrap	Character. One of "none" (the default, no bootstrapping), "numbag" (presence function is bootstrapped), or "doublebag" (presence and background functions are bootstrapped).
bootstrap_reps	Integer. Number of bootstrap replicates to use (default is 100)
quantile	Quantile to use for thresholding. Default is 0.05 (5 pct training presence). Set to 0 for minimum training presence (MTP).
constraint_regions	See <code>get_env_bg</code> documentation
background_buffer_width	Numeric or NULL. Width (meters or map units) of buffer to use to select background environment. If NULL, uses max dist between nearest occurrences.
standardize_preds	Logical. Should environmental layers be scaled? Default is TRUE.
...	Additional parameters passed to internal functions.

### Details

Current plug-and-play methods include: "gaussian", "kde", "vine", "rangebagging", "lobagoc", and "none". Current density ratio methods include: "ulsif", "rulsif".

### Value

A list containing 1) a data.frame containing cross-validated model performance statistics (`fold_results`), and 2) a data.frame containing model performance statistics evaluated on the full dataset (`overall_results`).

### Note

Either `method` or both `presence_method` and `background_method` must be supplied.

### Examples

```
{
# load in sample data

library(S4DM)
library(terra)

# occurrence points
data("sample_points")
occurrences <- sample_points
```

```

# environmental data
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))

# rescale the environmental data

env <- scale(env)

# Evaluate a gaussian/gaussian model calculated with the numbag approach
# using 10 bootstrap replicates.

evaluate_range_map(occurrences = occurrences,
  env = env,
  method = NULL,
  presence_method = "gaussian",
  background_method = "gaussian",
  bootstrap = "numbag",
  bootstrap_reps = 10,
  quantile = 0.05,
  constraint_regions = NULL,
  background_buffer_width = 100000)

}

```

---

fit\_density\_ratio      *Fit density-ratio distribution models in a plug-and-play framework.*

---

### Description

This function fits density-ratio species distribution models for the specified density-ratio method (Drake and Richards 2018; Maitner et al. 2026).

### Usage

```
fit_density_ratio(presence = NULL, background = NULL, method = NULL, ...)
```

### Arguments

presence	dataframe of covariates at presence points
background	Dataframe of covariates at background points
method	Character. See "notes" for options.
...	Additional parameters passed to internal functions.

### Details

Current methods include: "ulsif", "rulsif", "klien", "maxnet"

**Value**

List of class "dr\_model" containing model objects and metadata needed for projecting the fitted models.

**References**

Drake JM, Richards RL (2018). "Estimating environmental suitability." *Ecosphere*, **9**(9), e02373. [doi:10.1002/ecs2.2373](https://doi.org/10.1002/ecs2.2373).

Maitner BS, Richards RL, Carlson BS, Drake JM, Merow C (2026). "Flexible methods for species distribution modeling with small samples." *Ecography*, **2026**(2), e08112. [doi:10.1002/ecog.08112](https://doi.org/10.1002/ecog.08112).

**Examples**

```
# load in sample data

library(S4DM)
library(terra)

# occurrence points
data("sample_points")
occurrences <- sample_points

# environmental data
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))

# rescale the environmental data

env <- scale(env)

# Get presence environmental data

pres_env <- get_env_pres(coords = occurrences,
                        env = env)

# Get background environmental data

bg_env <- get_env_bg(coords = occurrences,
                    env = env,width = 100000)

# Note that the functions to get the environmental data return lists,
# and only the "env" element of these is used in the fit function

rulsif_fit <- fit_density_ratio(presence = pres_env$env,
                              background = bg_env$env,
                              method = "rulsif")
```

---

fit_plug_and_play	<i>Fit presence-background distribution models in a plug-and-play framework.</i>
-------------------	--

---

### Description

This function fits presence-background species distribution models for the specified plug-and-play methods (Drake and Richards 2018; Drake 2015; Maitner et al. 2026).

### Usage

```
fit_plug_and_play(
  presence = NULL,
  background = NULL,
  method = NULL,
  presence_method = NULL,
  background_method = NULL,
  bootstrap = "none",
  bootstrap_reps = 100,
  ...
)
```

### Arguments

presence	dataframe of covariates at presence points
background	Optional. Dataframe of covariates at background points
method	Optional. If supplied, both presence and background density estimation will use this method.
presence_method	Optional. Method for estimation of presence density.
background_method	Optional. Method for estimation of background density.
bootstrap	Character. One of "none" (the default, no bootstrapping), "numbag" (presence function is bootstrapped), or "doublebag" (presence and background functions are bootstrapped).
bootstrap_reps	Integer. Number of bootstrap replicates to use (default is 100)
...	Additional parameters passed to internal functions.

### Details

Current methods include: "gaussian", "kde", "vine", "rangebagging", "lobagoc", and "none".

### Value

List of class "pnp\_model" containing model objects and metadata needed for projecting the fitted models.

## Note

Either method or both presence\_method and background\_method must be supplied.

## References

Drake JM (2015). “Range bagging: a new method for ecological niche modelling from presence-only data.” *J. R. Soc. Interface*, **12**(107). doi:10.1098/rsif.2015.0086.

Drake JM, Richards RL (2018). “Estimating environmental suitability.” *Ecosphere*, **9**(9), e02373. doi:10.1002/ecs2.2373.

Maitner BS, Richards RL, Carlson BS, Drake JM, Merow C (2026). “Flexible methods for species distribution modeling with small samples.” *Ecography*, **2026**(2), e08112. doi:10.1002/ecog.08112.

## Examples

```
# load in sample data

library(S4DM)
library(terra)

# occurrence points
data("sample_points")
occurrences <- sample_points

# environmental data
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))

# rescale the environmental data

env <- scale(env)

# Get presence environmental data

pres_env <- get_env_pres(coords = occurrences,
                        env = env)

# Get background environmental data

bg_env <- get_env_bg(coords = occurrences,
                    env = env,width = 100000)

# Note that the functions to get the environmental data return lists,
# and only the "env" element of these is used in the fit function

kde_fit <- fit_plug_and_play (presence = pres_env$env,
                             background = bg_env$env,
                             method = "kde")
```

---

get_env_bg	<i>Extract background data for SDM fitting.</i>
------------	---

---

### Description

This function extracts background data around known presence records.

### Usage

```
get_env_bg(
  coords,
  env,
  method = "buffer",
  width = NULL,
  constraint_regions = NULL,
  standardize = TRUE
)
```

### Arguments

coords	Coordinates (long,lat) to extract values for
env	Environmental SpatRaster(s) in any projection
method	Methods for getting bg points. Current option is buffer
width	Numeric or NULL. Width (meters or map units) of buffer. If NULL, uses max dist between nearest occurrences.
constraint_regions	An optional spatialpolygons* object that can be used to limit the selection of background points.
standardize	Logical. If TRUE, the variables will be scaled and centered

### Value

A list containing 1) the background data (env), 2) the cell indices for which the background was taken (buffer\_cells), 3) the environmental means (env\_mean; NA if standardization not done), and 4) the environmental standard deviations (env\_sds; NA if standardization not done).

### Note

If supplying constraint\_regions, any polygons in which the occurrences fall are considered fair game for background selection. This background selection is, however, still limited by the buffer as well.

**Examples**

```
{  
  
# load in sample data  
  
library(S4DM)  
library(terra)  
  
# occurrence points  
data("sample_points")  
occurrences <- sample_points  
  
# environmental data  
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))  
  
# rescale the environmental data  
  
env <- scale(env)  
  
bg_data <- get_env_bg(coords = occurrences,  
                      env = env,  
                      method = "buffer",  
                      width = 100000)  
  
}
```

---

`get_env_pres`*Extract presence data for SDM fitting.*

---

**Description**

This function extracts presence data at known presence records.

**Usage**

```
get_env_pres(coords, env, env_bg = NULL)
```

**Arguments**

<code>coords</code>	Coordinates (long,lat) to extract values for
<code>env</code>	Environmental SpatRaster(s) in any projection
<code>env_bg</code>	Background data produced by <code>get_env_bg</code> , used for re-scaling

**Value**

A list containing 1) the environmental data at the presence locations (`env`), and 2) an `sf` data.frame containing the occurrence records(`occurrence_sf`).

## Examples

```
{  
  
# load in sample data  
  
library(S4DM)  
library(terra)  
  
# occurrence points  
data("sample_points")  
occurrences <- sample_points  
  
# environmental data  
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))  
  
# rescale the environmental data  
  
env <- scale(env)  
  
env_pres <- get_env_pres(coords = occurrences,  
                        env = env)  
  
}
```

---

get\_response\_curves     *Generate Response Curves*

---

## Description

Given an environmental data set, fitted models, and a directory to output plots, this function generates response curves for each predictor in the model. The response curves depict the predicted change in probability of presence as a function of the environmental predictor while holding all other predictors constant at their mean values.

## Usage

```
get_response_curves(  
  env_bg,  
  env_pres,  
  pnp_model,  
  n.int = 1000,  
  envMeans = NULL,  
  envSDs = NULL  
)
```

## Arguments

env\_bg                    Object returned by get\_env\_bg

env_pres	Object returned by get_env_pres
pnpl_model	Object returned by fit_plug_and_play or fit_density_ratio
n.int	Number of points along which to calculate the response curve
envMeans	A vector of means for each environmental predictor in the dataset. (not used)
envSDs	A vector of standard deviations for each environmental predictor in the dataset.(not used)

### Value

This function generates a set of marginal predictions for each environmental variable, holding other variables constant

### Author(s)

Cory Merow, modified by Brian Maitner

---

make_range_map	<i>Make a range map using plug-and-play modeling.</i>
----------------	---

---

### Description

This function produces range maps using plug-and-play modeling with either presence-background or density-ratio approaches.

### Usage

```
make_range_map(  
  occurrences,  
  env,  
  method = NULL,  
  presence_method = NULL,  
  background_method = NULL,  
  bootstrap = "none",  
  bootstrap_reps = 100,  
  quantile = 0.05,  
  background_buffer_width = NULL,  
  constraint_regions = NULL,  
  verbose = FALSE,  
  standardize_preds = TRUE,  
  ...  
)
```

**Arguments**

occurrences	Presence coordinates in long,lat format.
env	Environmental rasters
method	Optional. If supplied, both presence and background density estimation will use this method.
presence_method	Optional. Method for estimation of presence density.
background_method	Optional. Method for estimation of background density.
bootstrap	Character. One of "none" (the default, no bootstrapping), "numbag" (presence function is bootstrapped), or "doublebag" (presence and background functions are bootstrapped).
bootstrap_reps	Integer. Number of bootstrap replicates to use (default is 100)
quantile	Quantile to use for thresholding. Default is 0.05 (5 pct training presence). Set to 0 for minimum training presence (MTP), set to NULL to return continuous raster.
background_buffer_width	The width (in m for unprojected rasters and map units for projected rasters) of the buffer to use for background data. Defaults to NULL, which will take the maximum distance between occurrence records.
constraint_regions	See get_env_bg documentation
verbose	Logical. If TRUE, prints progress messages.
standardize_preds	Logical. Should environmental layers be scaled? Default is TRUE.
...	Additional parameters passed to internal functions.

**Details**

Current plug-and-play methods include: "gaussian", "kde","vine","rangebagging", "lobagoc", and "none". Current density ratio methods include: "ulsif", "rulsif",and "maxnet".

**Value**

A SpatRaster object containing a range map. Maps may be either binary or continuous, depending upon the quantile argument.

**Note**

Either method or both presence\_method and background\_method must be supplied.

**Examples**

```
{
# load in sample data
```

```
library(S4DM)
library(terra)

# occurrence points
data("sample_points")
occurrences <- sample_points

# environmental data
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))

# rescale the environmental data

env <- scale(env)

map <- make_range_map(occurrences = occurrences,
                      env = env,
                      method = "gaussian",
                      presence_method = NULL,
                      background_method = NULL,
                      bootstrap = "none",
                      bootstrap_reps = 100,
                      quantile = 0.05,
                      background_buffer_width = 100000)

plot(map)

}
```

---

project\_density\_ratio *Projects fitted density-ratio distribution models onto new covariates.*

---

### Description

This function projects fitted density-ratio species distribution models onto new covariates.

### Usage

```
project_density_ratio(dr_model, data)
```

### Arguments

dr_model	A fitted density ratio model produced by <code>fit_density_ratio</code>
data	covariate data

### Value

A vector of relative occurrence rates evaluated at the covariates supplied in the data object.

---

project\_plug\_and\_play *Projects fitted plug-and-play distribution models onto new covariates.*

---

### Description

This function projects fitted plug-and-play species distribution models onto new covariates.

### Usage

```
project_plug_and_play(pnp_model, data, bg_zero_ror_zero = TRUE)
```

### Arguments

pnp_model	A fitted plug-and-play model produced by fit_plug_and_play
data	covariate data
bg_zero_ror_zero	Logical. Default = TRUE. If the background is zero, set the ROR to zero. This will prevent NaN and Inf values which would otherwise interfere with AUC calculations.

### Value

A vector of relative occurrence rates evaluated at the covariates supplied in the data object.

### Note

The tsearchn function underlying rangebagging seems to fail sometimes with very uneven predictors. Rescaling helps.

---

sample\_points *Example S4DM occurrence data*

---

### Description

A sample dataset containing occurrence records.

### Usage

```
sample_points
```

### Format

A data.frame with 65 observations of 2 variables:

**Longitude** Longitude, in decimal degrees

**Latitude** Latitude, in decimal degrees ...

**Source**

<https://biendata.org>

---

sdm_threshold	<i>Thresholds a continuous relative occurrence rate raster to create a binary raster.</i>
---------------	---

---

**Description**

This function thresholds a continuous relative occurrence rate raster to produce a binary presence/absence raster.

**Usage**

```
sdm_threshold(  
  prediction_raster,  
  occurrence_sf,  
  quantile = 0.05,  
  return_binary = TRUE  
)
```

**Arguments**

prediction_raster	Raster containing continuous predictions of relative occurrence rate to be thresholded.
occurrence_sf	An sf object containing presence locations. Should be in the projection of the prediction raster
quantile	Numeric between 0 and 1. Quantile to use for thresholding (defaults to 0.05). Set to 0 for minimum training presence.
return_binary	LOGICAL. Should the raster returned be binary (presence/absence)? If FALSE, predicted presences will retain their "suitability" scores.

**Value**

A SpatRaster object containing a range map. Maps may be either binary or continuous, depending upon the return\_binary argument.

**Author(s)**

Cecina Babich Morrow (modified by Brian Maitner)

**Examples**

```
{  
  
# load in sample data  
  
library(S4DM)  
library(terra)  
  
# occurrence points  
data("sample_points")  
occurrences <- sample_points  
  
# environmental data  
env <- rast(system.file('ex/sample_env.tif', package="S4DM"))  
  
# rescale the environmental data  
  
env <- scale(env)  
  
bg_data <- get_env_bg(coords = occurrences,  
                     env = env,  
                     method = "buffer",  
                     width = 100000)  
  
pres_data <- get_env_pres(coords = occurrences,  
                          env = env)  
  
pnp_model <- fit_plug_and_play(presence = pres_data$env,  
                              background = bg_data$env,  
                              method = "gaussian")  
  
pnp_continuous <- project_plug_and_play(pnp_model = pnp_model,  
                                       data = bg_data$env)  
  
#Make an empty raster to populate  
out_raster <- env[[1]]  
values(out_raster) <- NA  
  
# use the bg_data for indexing  
out_raster[bg_data$bg_cells] <- pnp_continuous  
  
plot(out_raster)  
  
#convert to a binary raster  
  
out_raster_binary <-  
  sdm_threshold(prediction_raster = out_raster,  
                occurrence_sf = pres_data$occurrence_sf,  
                quantile = 0.05,  
                return_binary = TRUE)  
  
plot(out_raster_binary)
```

```
}
```

---

stratify_random	<i>Split data for k-fold spatially stratified cross validation</i>
-----------------	--

---

**Description**

Splitting tool for cross-validation

**Usage**

```
stratify_random(occurrence_sf, nfolds = NULL)
```

**Arguments**

occurrence\_sf a sf object containing occurrence records  
nfolds number of desired output folds.

**Details**

See Examples.

**Value**

Returns a sf dataframe containing fold designation for each point.

**Author(s)**

Cory Merow [cory.merow@gmail.com](mailto:cory.merow@gmail.com)

**Examples**

```
{  
  
# load in sample data  
  
library(S4DM)  
library(terra)  
library(sf)  
  
# occurrence points  
data("sample_points")  
occurrences <- sample_points  
  
occurrences <- st_as_sf(x = occurrences, coords = c(1,2))
```

```
random_folds <- stratify_random(occurrence_sf = occurrences,  
                               nfolders = 5)  
  
}
```

---

stratify\_spatial      *Split data for k-fold spatially stratified cross validation*

---

### Description

Splitting tool for cross-validation

### Usage

```
stratify_spatial(occurrence_sf, nfolders = NULL, nsubclusters = NULL)
```

### Arguments

`occurrence_sf` a sf object containing occurrence points

`nfolders` number of desired output folds. Default value of NULL makes a reasonable guess based on sample size.

`nsubclusters` intermediate number of clusters randomly split into nfolders. Default value of NULL makes a reasonable guess based on sample size. If you specify this manually, it should be an integer multiple of nfolders.

### Details

See Examples.

### Value

Returns a SpatialPoints dataframe with the data.frame containing fold designation for each point.

### Author(s)

Cory Merow [cory.merow@gmail.com](mailto:cory.merow@gmail.com)

### Examples

```
{  
  
# load in sample data  
  
library(S4DM)  
library(terra)  
library(sf)
```

```
# occurrence points
data("sample_points")
occurrences <- sample_points

occurrences <- st_as_sf(x = occurrences, coords = c(1,2))

manual <- stratify_spatial(occurrence_sf = occurrences, nfold = 5, nsubclusters = 5)
default <- stratify_spatial(occurrence_sf = occurrences)

}
```

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