# Package 'S4DM'

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Title Small Sample Size Species Distribution Modeling

Version 0.0.1

## **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 Drake and Richards (2018) <doi:10.1002/ecs2.2373>, Drake (2015) <doi:10.1098/rsif.2015.0086>, and 00202.1>.

```
Depends R (>= 3.5.0)
```

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**Encoding UTF-8** 

LazyData true

VignetteBuilder knitr

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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.

## 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)

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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.

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

# occurrence points
data("sample_points")
occurrences <- sample_points
# environmental data</pre>
```

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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.

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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.

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.

```
# 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</pre>
```

fit\_density\_ratio

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).

#### 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", "maxnet"

## Value

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

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#### References

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

#### **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"))</pre>
 # rescale the environmental data
  env <- scale(env)</pre>
 # Get presence environmental data
 pres_env <- get_env_pres(coords = occurrences,</pre>
                            env = env)
# Get background environmental data
bg_env <- get_env_bg(coords = occurrences,</pre>
                      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")
```

 $fit\_plug\_and\_play$ 

Fit presence-background distribution models in a plug-and-play framework.

#### **Description**

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

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#### 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). http://dx.doi.org/10.1098/rsif.2015.0086.

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

get\_env\_bg

#### **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"))</pre>
 # rescale the environmental data
   env <- scale(env)</pre>
# Get presence environmental data
 pres_env <- get_env_pres(coords = occurrences,</pre>
                            env = env)
# Get background environmental data
bg_env <- get_env_bg(coords = occurrences,</pre>
                       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$env,</pre>
                                 background = bg_env$env,
                                 method = "kde")
```

get\_env\_bg

Extract background data for SDM fitting.

## **Description**

This function extracts background data around known presence records.

#### Usage

```
get_env_bg(
  coords,
```

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```
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.

```
# 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)</pre>
```

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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

coords Coordinates (long,lat) to extract values for
env Environmental SpatRaster(s) in any projection
env\_bg Background data produced by get\_env\_bg, 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).

```
# 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)</pre>
```

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## 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
pnp_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

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make\_range\_map

Make a range map using plug-and-play modeling.

#### **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.

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```
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.

```
{
# 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"))</pre>
 # rescale the environmental data
   env <- scale(env)</pre>
   map <- make_range_map(occurrences = occurrences,</pre>
                          env = env,
                          method = "gaussian",
                          presence_method = NULL,
                          background_method = NULL,
                          bootstrap = "none",
```

project\_density\_ratio 15

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 fit\_density\_ratio

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)
```

#### **Arguments**

pnp\_model A fitted plug-and-play model produced by fit\_plug\_and\_play

data covariate data

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## 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

Thresholds a continuous relative occurrence rate raster to create a binary raster.

## **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
)
```

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## 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)

```
{
# 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"))</pre>
# rescale the environmental data
  env <- scale(env)</pre>
 bg_data <- get_env_bg(coords = occurrences,</pre>
                        env = env,
                         method = "buffer",
                         width = 100000)
 pres_data <- get_env_pres(coords = occurrences,</pre>
                             env = env)
 pnp_model <-fit_plug_and_play(presence = pres_data$env,</pre>
                    background = bg_data$env,
```

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```
method = "gaussian")
pnp_continuous <- project_plug_and_play(pnp_model = pnp_model,</pre>
                                           data = bg_data$env)
 #Make an empty raster to populate
 out_raster <- env[[1]]</pre>
 values(out_raster) <- NA</pre>
 # use the bg_data for indexing
out_raster[bg_data$bg_cells] <- pnp_continuous</pre>
 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

Split data for k-fold spatially stratified cross validation

#### 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.

stratify\_spatial 19

#### Author(s)

Cory Merow 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, nfolds = 5)
}</pre>
```

stratify\_spatial

Split data for k-fold spatially stratified cross validation

#### **Description**

Splitting tool for cross-validation

## Usage

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

## **Arguments**

occurrence\_sf a sf object containing occurrence points

nfolds 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 nfolds. Default value of

NULL makes a reasonable guess based on sample size. If you specify this man-

ually, it should be an integer multiple of nfolds.

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## **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

```
{
# 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, nfolds = 5, nsubclusters = 5)
default <- stratify_spatial(occurrence_sf = occurrences)
}</pre>
```

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